

Summary Sheet

If you have ever been on an aircraft, it is plain to see how slow boarding and disembarking is. For many this is insignificant, but for an airline company saving even a couple of minutes for each flight's boarding and disembarking will result in huge savings when considering the tens of thousands of airports and flights that occur each time. For an industry still struggling from the collapse of the tourism industry due to COVID-19, optimal and robust boarding and disembarking methods must be found.

To achieve this we developed two models, one for each of boarding and disembarking. As boarding and disembarking planes is an inherently stochastic process, we created a computational simulation over a pure mathematical model. Thus, we could better account for variable human behaviours and scenarios, giving a much more accurate distribution of data. Whilst many models already exist for this purpose, a key point of difference of our model is a greater consideration to several aspects of human behaviour. Namely, disobedience of boarding instructions, and travelling in groups.

We first modelled the Narrow Body Aircraft, simulating different boarding and disembarking methods using a Monte Carlo method. To create different boarding methods, we generated a randomized queue of passengers in the order that the boarding method prescribes (accounting for disobedient people) which could then be simulated boarding. Over many simulations, we could obtain an accurate average for the total time taken, allowing us to determine the most optimal method (least time taken). We also proposed two additional methods and ran them through the same simulations.

To simulate disembarking, we gave all seated passengers a priority value. Disembarking was carried out by moving passengers towards the exit at different rates dependent on their priority level. By altering the priority values we could carry out different disembarking methods and account for disobedience.

Both models implemented real-world data for factors such as moving speeds. This was to ensure the highest accuracy of our resulting times. We comprehensively analysed the results of these simulations, determining the effect of altering variables such as the number of people who disobey instructions, and varying numbers of carry-on baggage.

We adapted our models to two other passenger aircraft, the Flying Wing and the Two-Entrance Two-Aisle, and applied the most optimal boarding and disembarking methods used on the Narrow Body plane. Furthermore, we considered the effect of a reduced capacity of the passenger aircraft, a relevant deliberation in the age of COVID-19.

Overall, it was found that for boarding, one of our own proposed methods – boarding in the order of window, middle and aisle seats with the allowance of groups to board together – was on the whole the most optimal over the three aircraft. The optimal disembarking method was one in which the plane was unloaded from the back of the craft to the front.

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1 Introduction

1.1 Background

As society becomes increasingly globalised, the importance of air travel grows. Flight numbers before the COVID-19 pandemic were at an all-time high, and they have doubled in the past 20 years. Following a temporary disruption due to COVID-19, this trend appears ready to continue its steady upwards climb[1]. This has the consequence that small optimisation changes can result in enormous savings for both airline companies, passengers, and airports in terms of usable time wasted. Some of the biggest bottlenecks for plane turnaround are boarding and disembarking efficiency - that is, the way that passengers are loaded to and unloaded from planes[2]. There exist a variety of methods for these processes, each with varying theoretical and practical efficacies. As such, this report presents our developed model and simulates different onboarding and embarking methods for various aircraft models.

1.2 Problem Restatement

To ascertain the efficiency of different systems, we will develop two models with allowances for practical considerations that can be adapted to a variety of conditions.

1. Develop a plane boarding model and disembarking model which allows us to test the efficiency of different boarding/disembarking methods on a narrow-body plane
2. Adapt the models to test on different aircraft types (i.e. Flying Wing and Two-Entrance Two-Aisle) and also the effects of limited capacity flights due to COVID-19
3. Write a one-page letter to an airline executive that explains our results and its benefits to their airline

1.3 Basic Assumptions

Our initial model uses a few basic assumptions. The aircraft is to be divided into cells which one person can occupy at a time. The aisle space between rows and each seat is represented by one cell.

- Only one person can comfortably walk in an aisle cell
Justification: Although aisle width varies by aircraft, a reasonable estimate is 0.50m wide[3]. On average, men have longer shoulder width than women, at 0.41m wide[4] and passengers are often carrying luggage which increases their width requirement. Thus, it is reasonable to assume that only one person can walk down the aisle at a time, with passengers both being laden with bags, respecting personal space, and potentially being weary of close contact due to infection risks. As such, when a passenger is loading their carry-on luggage into an overhead bin, the aisle is also blocked.
- Seated passengers block passengers who wish to sit further down in the same row
Justification: The passenger cannot leap over the seated passenger. Not only is this valid from a social etiquette perspective, but in the provided aircraft designs, legroom looks to be minimal so it is physically unfeasible too.
- When a seat passenger leaves a row to make room for an incoming passenger, they are momentarily able to inhabit the same aisle cell
Justification: As the passenger will want to reach their seat, they will not mind temporarily having reduced room as they move into their seat cell.

- Time to walk one aisle cell is constant

Justification: This time was obtained by analysing a sample of $n = 10$ YouTube videos of people walking down aisles on flights, by counting the number of frames elapsed when each individual walks one aisle cell, and the playback details of the YouTube videos (typically either 60 or 30 frames per second - these are listed in the references). Using this, we can determine that the time to move one cell down the aisle is given by $1.05s$.

1.4 Variables and Factors

Several variables were used in our model to account for real-life phenomena. Some of these will be expanded on in later sections.

A **bag coefficient** was used to give a weighted probability of each passenger having carry-on luggage that they would want to stow in an overhead locker.

Another variable was the **number of groups**. Passenger populations are not homogenous; often they contain inseparable groups such as families of varying sizes. Members of these groups were seated adjacently in the same row and entered the plane in adjacent cells too. Upon entering the plane, it was assumed that groups would be in an order that would minimize blockage when getting into seats (i.e. in the order window, middle, aisle). This is reasonable as groups would want to minimize their own inconvenience and could communicate with each other to align themselves in this order. This factor has an appreciable effect on different boarding methods and was rarely investigated with any depth in any of the papers found in our literature review.

A **disobedience coefficient** was introduced to model the common scenario of passengers not following instructions. In these cases, a passenger (or group) would enter the plane in a different boarding category than ordered, which could be caused by ignorance, impatience or lateness. This, much like the **number of groups**, was rarely considered in an in-depth manner in the existing literature but would still significantly affect boarding times.

2 Narrow-Body Boarding

For both our models, we simulated the entire boarding/disembarking process. Keeping track of time during this simulation, we could calculate total boarding/disembarking time. Python 3.9 was used for this simulation.

2.1 Boarding Model Situation

To model boarding, we designed an algorithm that would see all passengers make their way to their assigned seat. Once on the plane after waiting in the boarding queue, passengers would follow a rigid set of rules, and variation would naturally occur due to variation in input: passengers had randomly generated differing numbers of baggage, and orders in which they entered the plane. Different boarding methods would be accounted for in the order of which passengers in prioritized seats entered the plane. A simplified process of the model as experienced by a passenger is best represented in the flow chart in Fig 2.1. This logic is easily followed and provides a robust algorithm that passengers can follow.

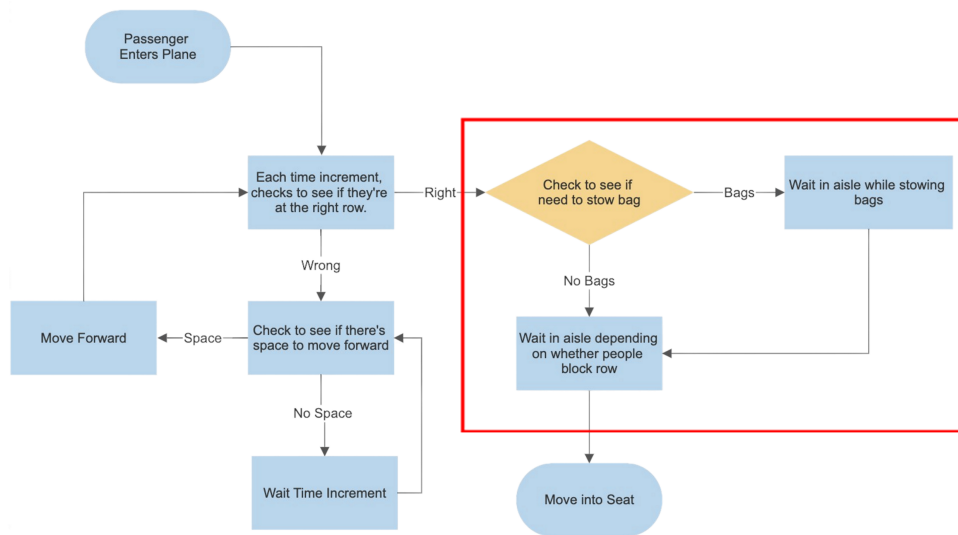


Figure 2.1: The logic behind passenger movement in the narrow-body aircraft

In the model this is simulated for all passengers simultaneously, as any passenger in the aisle could be at any step at any time. This is done by repetitively iterating down the aisle, starting from the passenger furthest from the entrance. Their state is determined, and an action done accordingly. Since it is assumed that there is a steady flow rate into the plane, if the first position in the aisle is ever empty, then the next passenger in the boarding queue occupies this space – ‘passenger enters plane’ in the flow chart. A key part of this simulation is the concept of an **internal clock**. Each passenger has this attribute, which counts down the real time (e.g., 1 sec) until they can complete an action. For example, the time to progress one cell forward is constant. The section of the flow chart enclosed in red is implemented in the simulation by calculating the total time that these actions would take and increasing the passenger’s internal clock until this time is achieved, whereupon they can undertake their action. A visualisation tool was used on the code, allowing us to generate real-time visualisations of the simulations (see Fig 2.2, and the code in Appendix NUMBERHERE).

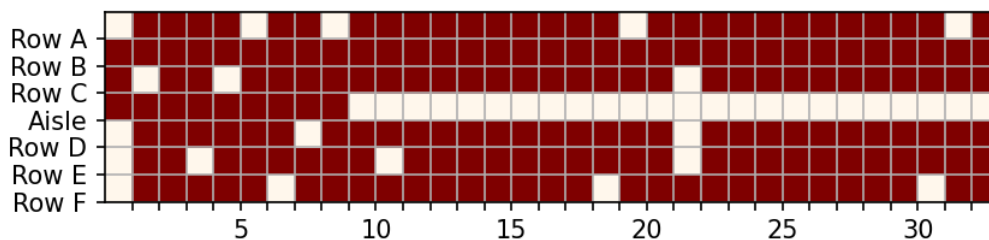


Figure 2.2: Visualisation tool in use on the narrow aircraft. Note that the aisle is currently blocked by a passenger in row 9.

In the following sections, we derive how these times are calculated.

2.1.1 Carry-on Baggage Delay

In airplanes it is commonplace that passengers load their carry-on luggage into the overhead bins. The aisle is blocked for the duration of this process. To account for this, the following piecewise function was developed to model the time that each passenger blocks the aisle while loading carry-on bags (which impedes the flow of passengers down the aisle).

Variable	Description
T_{bags}	Time that the main aisle is blocked due to carry-on luggage loading
n_{bags}	Number of carry-on bags to be stored in the overhead bins
n_{bins}	Number of carry-on bags present in overhead bins before storing
n_{max}	Maximum number of bags overhead bins can hold
C_0, C_1, C_2	Scaling constants depending on the value of n_{max}

$$T_{bags}(n_{bags}, n_{bins}, n_{max}) = \begin{cases} 0 & \text{if } n_{bags} = 0 \\ \frac{C_0}{1 - C_1 n_{bins}/n_{max}} & \text{if } n_{bags} = 1 \\ \frac{C_0}{1 - C_1 n_{bins}/n_{max}} + \frac{C_2}{1 - (n_{bins} + 1)/n_{max}} & \text{if } n_{bags} = 2 \end{cases} \quad (1)$$

The model only considers $n_{bags} \in \{0, 1, 2\}$ since it is assumed that the maximum number of carry-on items that each passenger is permitted to have is $n_{bags} = 2$. Many airlines, including Air New Zealand[5], impose this maximum (even for business class passengers). The benefit of this equation is in its generality; its many parameters allow for precise calibration to produce more accurate results, especially for different aircraft models. For the purposes of modelling the narrow plane, we assumed each row of three had an overhead bin with capacity $n_{max} = 6$ since each passenger in the row could carry at most $n_{bags} = 2$. This is assuming not all the stowed items are full size suitcases: some carry-ons are likely to be smaller items such as handbags/tote bags. The passengers will be able to fit more of these into an overhead bin, thus the larger capacity. Then, taking $n_{max} = 6$, the values of C_0, C_1, C_2 were calibrated to be 4, 0.8, and 2.25 respectively. This yields the following equation, which was implemented into our model.

$$T_{bags}(n_{bags}, n_{bins}, 6) = \begin{cases} 0 & \text{if } n_{bags} = 0 \\ \frac{4}{1 - 0.8n_{bins}/6} & \text{if } n_{bags} = 1 \\ \frac{4}{1 - 0.8n_{bins}/6} + \frac{2.25}{1 - (n_{bins} + 1)/6} & \text{if } n_{bags} = 2 \end{cases} \quad (2)$$

The function is piecewise to easily account for the varying number of bags that each passenger carries. Passengers carrying no bags do not take time to stow, while those stowing two bags take longer than those stowing one bag (thus the added term). Another consideration is that the function is designed to increase when there is less space in the overhead bin (i.e., when n_{bins}/n_{max} is large) as passengers will have to find space and squeeze their bags in, increasing aisle blockage time. For instance, if a passenger has one bag and there are already 1/6 bags in the overhead bin, then $T_{bags}(1, 1, 6) = 4.6$. However, if the compartment is almost full with 5/6 bags, then $T_{bags}(1, 5, 6) = 12$ as the passenger will have to locate a space and squeeze their carry-on in.

2.1.2 Shifting Seats Delay

Another large source of aisle blockage arises from the common situation where a passenger tries to reach their seat in a row but is blocked by a seated passenger. Before the passenger can reach their seat, the seated individual must stand up and move out to the aisle to allow the passenger to reach their seat, before sliding back. This process is lengthy and will impede the flow of passengers down the main aisle. This is furthermore complicated by the fact that there are many variations on this scenario, with different seated passenger positions and passenger seat goals, which will have appreciably different delay times. To model the additional time needed for these different shuffles, Eq. 3 was derived.

Variable	Description
$T_{shuffle}$	Time that the main aisle is blocked
t_{up}	Time taken for a seated passenger to stand up
t_s	Time taken for a passenger to travel the width of a seat
f	The index of the furthest seat that blocks the passenger's seat
n_s	The number of seated passengers that block the passenger's seat

Let the seats be indexed such that the aisle seat has index 1 and the index of each consecutive seat increases until the window seat. Since we are only concerned with total time that the aisle is blocked, only the time that passengers are occupying the aisle needs to be kept track of. First, the person seated furthest from the aisle stands and moves into the aisle ($t_{up} + ft_s$). Then the passenger moves into the row (t_s), and finally the previously seated passengers move back into the row (n_st_s).

$$\begin{aligned} T_{shuffle}(f, n_s) &= t_{up} + ft_s + t_s + n_st_s \\ &= t_{up} + t_s(f + 1 + n_s) \end{aligned} \quad (3)$$

Following this derivation, we state that the equation makes the following assumptions:

- The seated passengers notice the passenger once they are standing next to the row
- All the required seated passengers stand up at the same time and begin to exit the row
- That two people can inhabit the aisle cell adjacent to the row (assumed earlier)
- Once the passenger has entered the row, the previously seated passengers begin moving back into the aisle, following right behind the passenger in the correct order

These assumptions are sufficiently realistic to generate results which closely model reality.

2.2 Boarding Queue Generation

A queue of passengers with assigned seats was generated to move into the aisle. By altering the order of the passengers in this queue, we could simulate different boarding methods. For example, we could place everyone in the queue in order of aft, middle, front. Within these sub-sections of the queue, the order was randomized each trial to further increase realism. At this point, we also assigned each passenger a discrete number of baggage, either 0, 1, or 2. This was done by utilizing a weighted probability. Overall, we implemented algorithms to create boarding queues for all the required boarding methods, as well as several others. However, to increase realism of the model, we added additional variation within these.

2.2.1 Disobedience Coefficient

Undoubtedly, there will be passengers who do not follow the rules of whichever boarding method is in place. This is due to two main reasons: impatience (boarding before they are called), and lateness (being late to their boarding time). These passengers are rarely accounted for in the literature, yet they have an appreciable effect on boarding times. To include this in our model, we introduced the **disobedience coefficient**, ψ , the probability of any passenger in the queue to not follow the desired boarding method. For instance, in a sectional boarding method, a passenger sitting in the aft section of the plane would have a ψ chance of boarding with a different group (and given that they do, a 50% chance for either group). Initially this was fixed at $\psi = 0.3$; online studies found that 30% of passengers are late for their flights, and we thought that this was a reasonable number that would be impatient as well.

2.2.2 Groups of People

Another important consideration in the model is the existence of groups of people that board together. Families, couples, and the like are present in high concentration on flights and are often seated together. Importantly and as discussed previously, they board together and enter the queue in the way in which they would enter seat rows, decreasing total boarding time. To account for this in our model, when a passenger in queue is generated, there is a weighted probability that they will be in a group of 1, 2 or 3. Groups of 1 are simply regular passengers. Groups of 2 or 3 are adjacent in the boarding queue and are seated in adjacent cells. Groups of 4 or larger were excluded since the aircraft only allowed a maximum of 3 to sit together in a row, effectively meaning a group above 3 can be split into two groups. Initially, the weighted probabilities of a passenger being in a group of 1, 2 or 3 was set at (20,80,10).

We also considered the effect of the disobedience coefficient on groups. We initially considered a group to be disobedient if any members of the group of size n were disobedient. However, as $(1 - \psi)$ is the probability that a passenger is obedient, then $(1 - \psi)^n$ is the probability that the entire group is obedient. Hence, $1 - (1 - \psi)^n$ is the probability that the group would be disobedient. For a ψ value of 0.3, this would create a disobedience probability of 0.51 for groups of 2 and 0.657 for groups of 3. We thought that this was unrealistically high, and instead determined that the disobedience probability would be ψ for the entire group.

2.2.3 Bag Coefficient

A key stochastic variable in this model is the number of carry-on bags that any given passenger will stow in the overhead lockers. Just as in real plane boarding, this is clearly prone to variation. To account for this, we introduced another 3-tuple in the code to give a weighted probability of a passenger stowing either 0, 1 or 2 bags. Unfortunately, there was a lack of available data on average passenger bag count online. As such, further analysis of the previous YouTube videos allowed us to tentatively obtain an estimate of (20,80,10). However, in the sensitivity analyses later this value was changed appropriately, allowing us to determine the validity of this initial assumption.

2.3 Modelled Results for Provided Boarding Methods

The three provided methods for boarding were random boarding, boarding by section, and boarding by seat. It was assumed that boarding by seat would make no allowances for groups of people. However, the other methods were modelled using groups.

2.3.1 Random Boarding

At first glance, the method of random boarding seems crude and inefficient. However, simulations run on our model reveal that the random method is reasonably effective. It took on average 689.4 seconds to finish boarding the plane, with a 5th percentile of 626.7s and a 95th percentile of 755.7s. This means that 90% of the values fall in this range of 129 seconds.

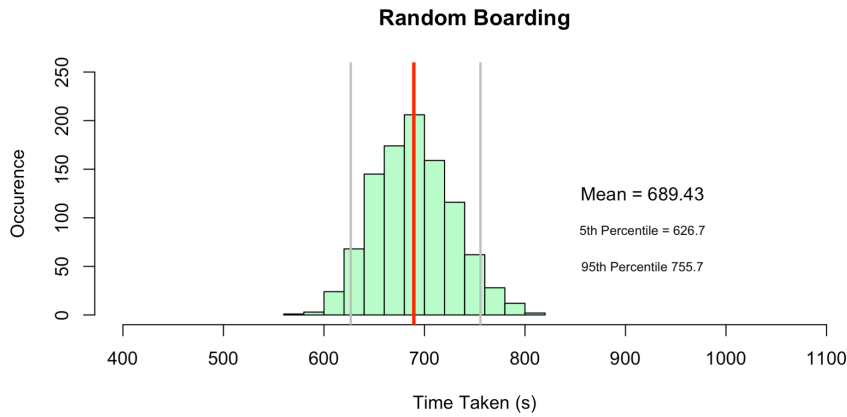


Figure 2.3: Monte Carlo simulation graph of the random boarding method

2.3.2 Boarding by Section

The second supplied method was to board the plane in sections. Boarding by aft (rows 23-33), middle (12-22) and front (rows 1-11) sections in varying order produced different results in our model. A set of results for all possible variations can be seen in the bar chart in Fig 6.1, but we discuss only the most and least optimal methods.

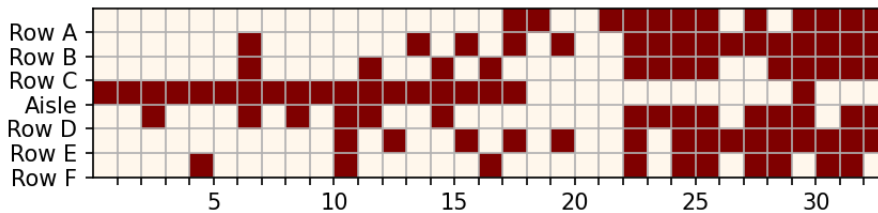


Figure 2.4: Visualised boarding by section starting with the aft. Note the disobedient passengers who have already seated themselves in the front and middle sections.

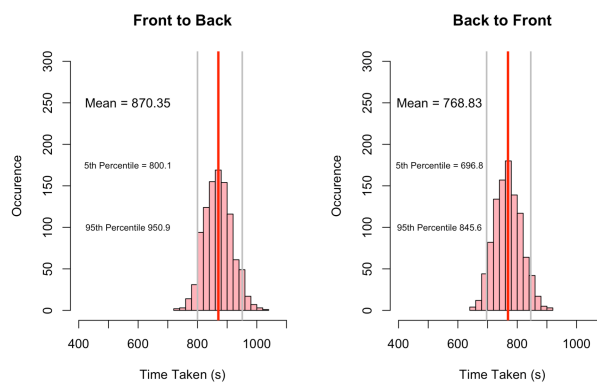


Figure 2.5: Monte Carlo simulation graphs of boarding front, middle, aft and aft, middle front

After running 10,000 trials, we found that the most optimal order of boarding was aft, middle, front (see breakdown in Appendix A). The mean time taken to fill up the narrow body airplane was **768.8 seconds**, with 90% of the times falling between 696.8s and 845.6s (spread of 148.8s). In comparison to this, it took on average **870.4 seconds** to board using the front, middle, aft method, with 90% of the times between 800.1s and 950.9s (spread of 150.8 seconds). This difference can be explained by considering Fig 2.6. On the left visualisation, the back fills first and so there is room to queue in the aisle, while on the right when the front is boarded first, the queue extends outside of the plane. Interestingly, this common method for boarding the plane is actually significantly slower than a random boarding order. However, the ability to simplistically split boarding into groups of people is valuable for airline companies, as it provides structure as to who should line up when. In the random boarding method, everyone is called to line up at once. This may potentially cause large queues and waste passengers time queuing in a long line.

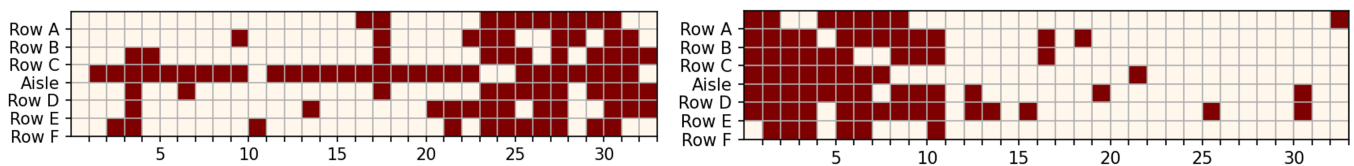


Figure 2.6: Visualisation of boarding by section, with AMF on the left and FMA on the right. Note the disobedience passengers sitting in the incorrect sections.

2.3.3 Boarding by Seat (WMA/WilMA)

The plane can also be boarded by seat type. This method allows all passengers with a window seat to board first, then middle, and finally aisle seats. Initially it seems like an ideal boarding method as it is relatively fast, with a mean boarding time 519.1 seconds. It is consistent too, with 90% of the values within 85 seconds of each other (5th percentile 479.1s, 95th percentile 564.1s). Not only this, but it is also straightforward to implement, with 3 easily definable groups of passengers. However, it splits groups. This is effectively unworkable in practice due to the separation of groups, particularly in the case of children and elderly.

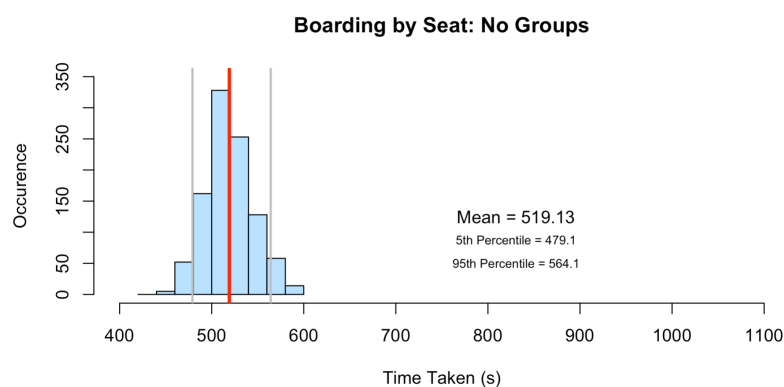


Figure 2.7: Monte Carlo simulation graph of boarding by seat without groups

2.3.4 Sensitivity Analysis of Provided Boarding Methods

We now perform a sensitivity analysis on the provided boarding methods.

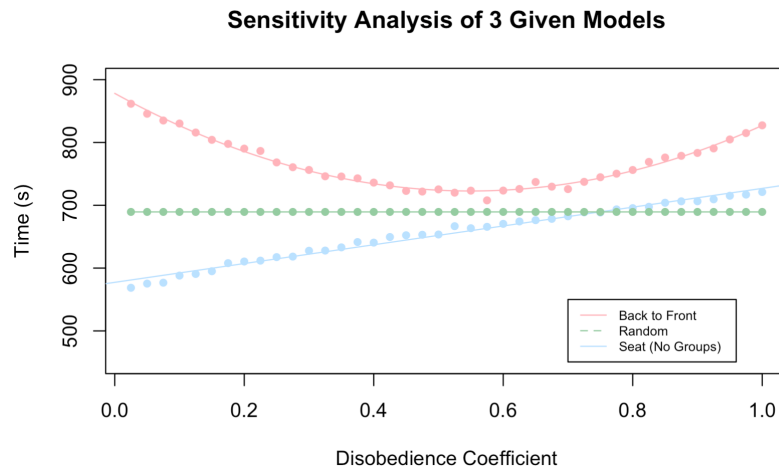


Figure 2.8: Sensitivity analysis of the disobedience coefficient on the interval $0 \leq \psi \leq 1$

Fig 2.8 shows the impact of changing the disobedience coefficient on the time taken to board, for the three given models. The effect of changing the disobedience coefficient for the section boarding was most interesting. As the number of people not following the prescribed method increased, the boarding method trended towards random. This meant the time taken decreased as random boarding is faster than section boarding. At a disobedience coefficient of $\psi = 0.5$, the boarding method is effectively random, thus the times are equivalent. However, as more people decide not to board with their prescribed group, the time starts to increase again. This is due to the boarding becoming 'ordered' again by section, which is slower than a random boarding method. This behaviour from the boarding by section method is ideal for airline companies, as a realistic extent of disobedience will help their boarding times. The random boarding method is completely insensitive to changes in disobedience, as there are no rules to disobey. The boarding by seat method without groups is the fastest boarding method provided, but it is also the method most impacted by changes in disobedience. This is potentially undesirable behaviour in a boarding method for airline companies, however under all reasonable values of the disobedience coefficient, boarding by seat is the fastest boarding method.

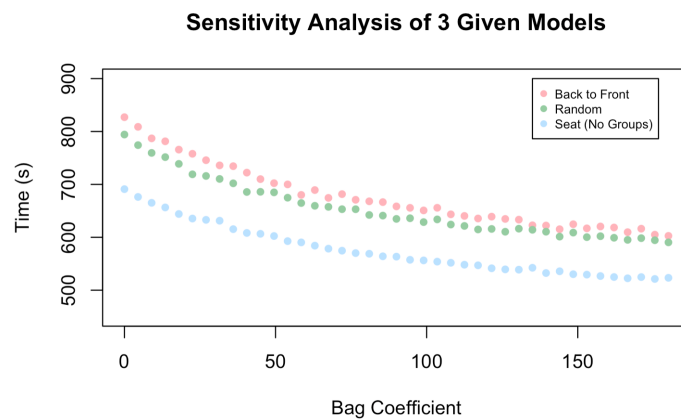


Figure 2.9: Sensitivity analysis of provided boarding methods by scaling a part of the bag coefficient.

Changing the bag coefficient changes the number of people without bags. The higher the coefficient, the higher the number of people without bags. The relevant time relating to bag numbers is the time spent in the aisle stowing. As such, only the number of bags stowed is pertinent to this model. Therefore, our variation of the bag coefficient (integers from 0 to 180) is effective at describing the impact of all plausible variations in bag numbers and bag stowage on the time take to board an

aircraft. From this analysis, we found that the three recommended methods are of equal sensitivity to variations in the bag coefficient. This is shown by the identical shape of the curves.

2.4 Modelled Results for Other Boarding Methods

2.4.1 Modified Steffen Method

The Steffen method is a plane boarding method proposed by Jason Steffen in 2008 which is suggested to be the method that produces the optimal plane boarding time[6]. However, this method is highly theoretical. It relies on the unrealistic assumption that passengers are efficient and highly organised. Instead, we present the modified version of the Steffen method which has a slightly larger grounding in reality. This method boards even numbered rows on the right hand side, then even rows on the left, then odd rows on the right, to odd rows on the left. This was almost the fastest boarding method we tested, with a mean time to board of 647.05 seconds. The 5th percentile was 595.3 seconds, and the 95th percentile was 696.5 seconds (a spread of 101.2 seconds).

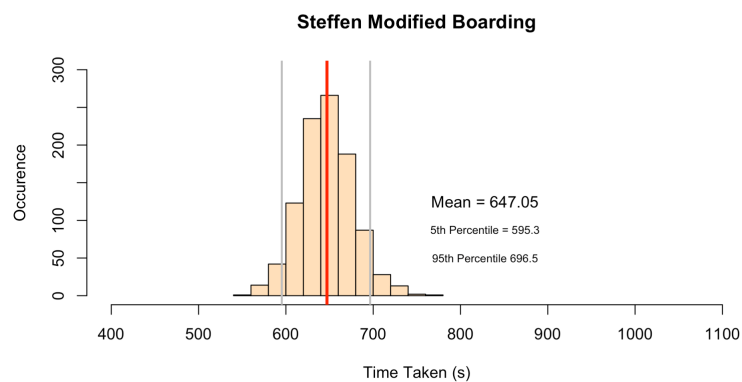


Figure 2.10: Monte Carlo simulation graph of the modified Steffen method

2.4.2 Prioritised Groups

In this method, passengers are classified as having one of two classes of walking speeds: normal and slow. This removes the need for our initial assumption that walking speed is relatively constant and allows us to test the validity of this assumption. Many airlines allow prioritised groups such as families with young children, disabled and elderly people to board first. The passengers in these prioritised groups are classified as having slow walking speed. We run this method through our model to determine its efficacy.

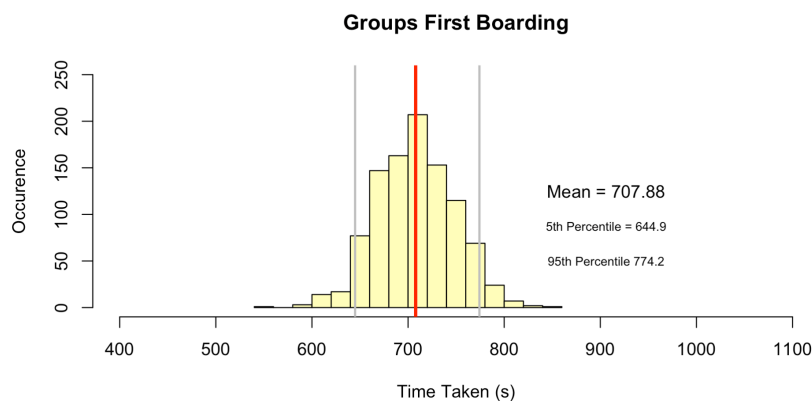


Figure 2.11: Monte Carlo simulation graph of the prioritised group boarding method

2.4.3 Modified Boarding by Seats (WMA)

As mentioned, the WMA has some serious drawbacks, particularly in regard to the splitting of groups. To overcome this, we devised a modified WMA method, which is one of our additional boarding methods. In this seating method, window seats are boarded first. However, if someone with a window seat is also part of a group, that whole group will board. The same thing occurs for middle seats and aisle seats. This avoids the problem of splitting groups while maintaining some of the efficiency of the WMA method.

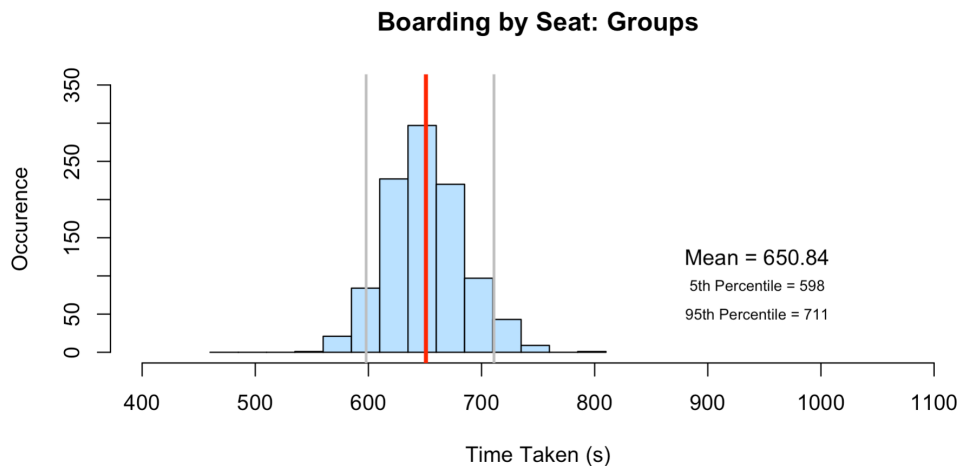


Figure 2.12: Monte Carlo simulation graph of the modified boarding by seats (WMA) method

The mean boarding time we obtained from this method was 650.84 seconds, with a 5th percentile of 598s and a 95th percentile of 711s (spread of 113 seconds). This adjusted method is relatively novel and hasn't seen much discussion in literature. However, its unique combination of practical and theoretical efficiency makes it an attractive proposition.

2.5 Optimal Boarding Method

After analysis of the previous five methods, we conclude that the modified WMA method is the best. The mean time to board the narrow body plane after 1,000 trials is 650.84 seconds. It should be noted that this isn't the optimal time that was achieved; the modified Steffen took only 647.05 seconds, and WMA without groups took 519.13s. This data is summarised in Fig 2.13. However, the modified WMA is significantly more practical to implement than both. The modified Steffen requires an unrealistic degree of coordination from random passengers and WMA without groups has the unrealistic assumption of splitting families and other groups apart. The modified WMA method allows for groups and can be easily implemented by airlines (by just calling seat letters to board, including family groups). It is also less sensitive to changes in the disobedience coefficient than alternative methods, such the Steffen modified. Although the time to board is initially slightly faster in the Steffen modified, as the disobedience coefficient increases, the time to board from the Steffen method increases faster than the time to board from the modified WMA. This is advantageous, as it means there is likely less variation in this modified WMA model in comparison to similarly fast boarding methods, allowing airline companies to better predict the boarding times.

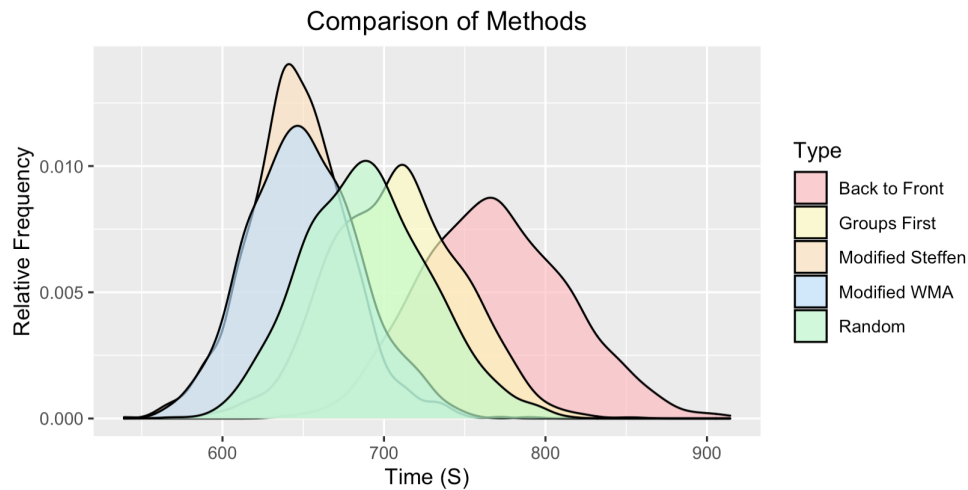


Figure 2.13: Comparison of Monte Carlo simulation graphs of different boarding models featured in previous sections

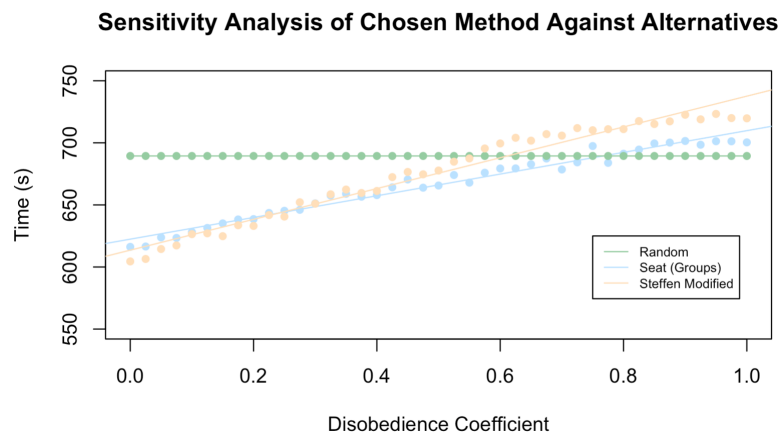


Figure 2.14: Sensitivity analysis of chosen methods for disobedience coefficient

3 Narrow-Body Disembarking

Having run simulations on our model under different boarding methods, we now turn our attention to the problem of disembarking. When exiting a plane, people typically move towards the nearest exit whenever space becomes available. A simulation of this is the basis of our disembarking model. By modelling individual interactions, such as what happens when two people come into the same space, we were able to ensure that our model was true as possible to a real disembarking.

3.1 Generation of Priority Map

The disembarking model runs through the generation of a priority map. Each person/group is assigned a priority value, representing how much they want to leave the plane. This is realistic since some people are desperate to leave and others being happy to sit on the plane until the rush dies down. This value is used when there is a passenger interaction. The priority values of each passenger that can move into the square are compared, and the passenger with highest priority is given the right of way. This map can also be manipulated to get different disembarking methods. By giving the highest priority to passengers we want to leave first, we can manipulate the order of who leaves first to find an optimal disembarking method. As such, different methods call for different priority

maps. The creation of the priority maps begins with the creation of an ideal priority map. In this map everyone would be assigned values such that they'd disembark in the desired fashion. Fig 3.1 shows an ideal priority map for disembarking by row, back to front, in the narrow body aircraft.

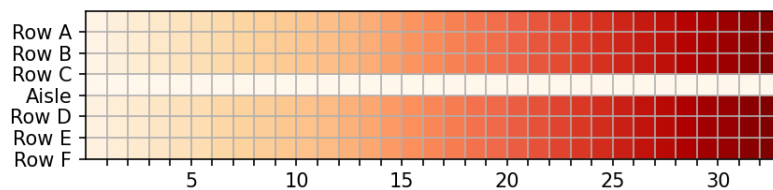


Figure 3.1: Ideal priority heatmap of back to front disembarking

In practice it is highly unlikely that everyone would follow a perfect disembarking model and therefore a disobedience coefficient was implemented, similar to the boarding model. The value of the disobedience coefficient was increased from 0.3 (in the boarding model) to 0.4 in this disembarking model. This choice was based on the fact that people are more likely to be tired, and may just want to leave the plane as soon as possible following a long flight. There is no feasible way to obtain data for this particular coefficient, and to investigate the effect this coefficient has on boarding times we performed a sensitivity analysis, varying the disobedience coefficient. Like in the boarding model, the disobedience coefficient describes the chance that a particular person won't follow their prescribed disembarking method. These disobedient people are then randomly assigned a new priority value ranging from 1 to the maximum possible priority value which varies depending on method. An implementation of this on the previously given priority map can be seen in Fig 3.2.

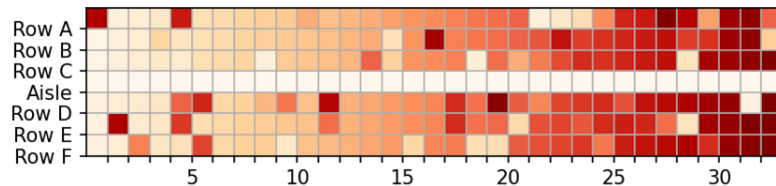


Figure 3.2: Introduction of disobedience coefficient to the ideal heatmap in Fig 3.1

As in the boarding method, we accounted for the fact that many people travel in groups that cannot be split. To implement this in the model, the priority of a group of size n is set to the mean of each member's priority in that group like so: $P_{group} = \frac{1}{n} \sum_{i=1}^n P_i$. The effect of this can be seen in Fig 3.3. Note the group in row 32 (seats ABC).

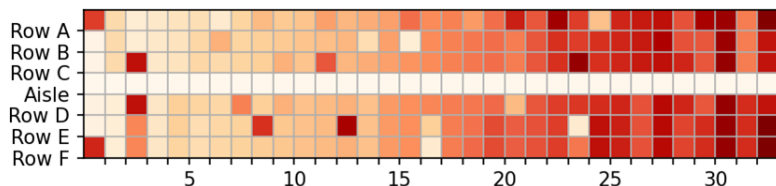


Figure 3.3: Introduction of groups to the heatmap in Fig 3.3

3.2 Logic of Disembarking

The diagram to the right shows the logic of the disembarking. By looping through the unoccupied aisle spots, and moving individuals into them, we can simulate the whole moving out. We considered the movement in and out of aisles as well.

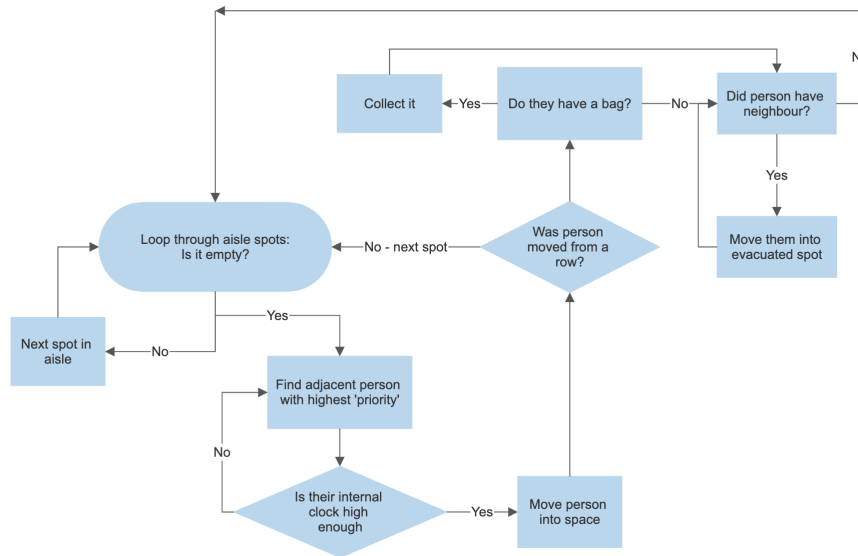


Figure 3.4: A flow diagram of the disembarking algorithm from a passenger's perspective

3.3 Time to Unstow Bags

Just as stowing bags during boarding blocks the aisle, the act of unstowing bags during disembarking blocks the aisle too. The following formula is a variant of Eq 1 that simply changes $n'_{bins} = n_{bins} - 2$. This is done to avoid division by zero, since many overhead bins will have $n_{bags} = 6$ as they are full. Eq 1 accounts for the number of bags already in bins – it takes longer difficult to remove a bag out of a packed luggage bin than an empty one. Note this n'_{bins} is simply labelled n_{bins} in Eq 1.

$$T_{bags}(n_{bags}, n_{bins}, 6) = \begin{cases} 0 & \text{if } n_{bags} = 0 \\ \frac{4}{1 - 0.8(n_{bins} - 1)/6} & \text{if } n_{bags} = 1 \\ \frac{4}{1 - 0.8(n_{bins} - 2)/6} + \frac{2.25}{1 - (n_{bins} - 1)/6} & \text{if } n_{bags} = 2 \end{cases} \quad (4)$$

3.4 Optimal Disembarking Method

The optimal disembarking method for the narrow body aircraft was found to be disembarking from back to front by row. This was initially surprising. However further analysis suggested it to be the quickest due to it having the greatest aisle flow out of all methods. The rate of free aisle flow hindered by retrieving baggage determines the rate people can enter the aisles and hence leave the plane. Back to front results in the greatest aisle flow due to people feeding into the aisles from the back of the plane. Should they need to retrieve a bag, they **a)** hold very few people up as they are at near the end of the queue, and therefore hold very few people up and **b)** by them stopping, they allow people in front of them flow into the queue meaning no gaps are left open.

This is opposite to the 'front to back' boarding method which is employed by most airlines and is the slowest disembarking method. This is because when someone enters the aisle from the front

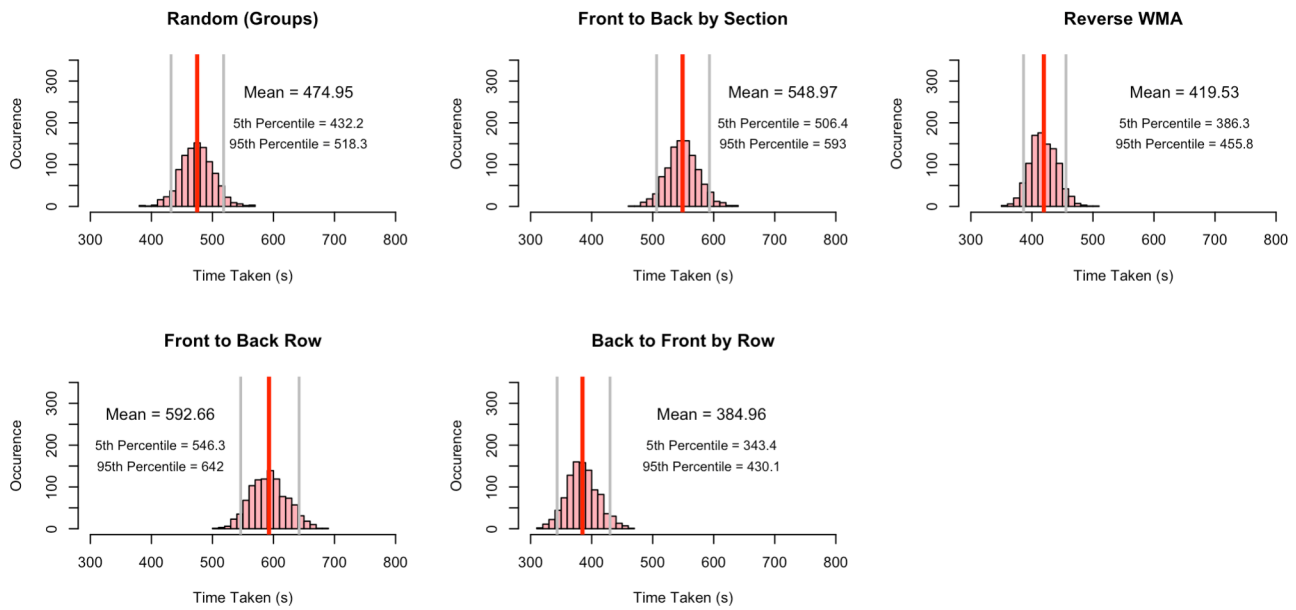


Figure 3.5: Monte Carlo simulation graphs of various disembarking methods (1k trials)

of the aircraft and retrieve their bag they hold the whole queue up whilst not allowing anyone in front of them to enter the queue as there is no one else in front. This back to front system would rely on a ‘right of way’ approach to disembarking the plane, where people at the back have the highest priority. When there is a space that two people could move into, the passenger at the front would have to give way to the passenger coming from behind them.

Other notable disembarking methods were the Reversed WMA which grants priority to aisle-seat passengers followed by middle then window seat. The Reversed WMA produces slower time than Back to Front as more people enter the queue near the front of the aircraft and thus block the queue as they retrieve bags. Reversed WMA is also impractical to implement as it requires a large degree of coordination in comparison to the relatively simple Back to Front method which is a reverse of the commonly used Front to Back Disembarking.

Disobedience Coefficient	Reverse WMA	Back to Front	Random
0	393	231	474.3
0.2	421.7	355	474.3
0.4	438.2	404.2	476.1
0.6	459.4	435.7	474.9
0.8	470.2	451	474.2
1.0	474.9	474.1	474.7

Table 1: Sensitivity analysis of disembarking methods by scaling the disobedience coefficient

This table shows that as disobedience increases, the time taken to disembark decreases. At no disobedience, we get fast disembarking times for reverse Wilma and back to front and at the maximum disobedience we see the times being similar to a random boarding time. Importantly this table also reveals that these models are very sensitive to disobedience especially back to front between 0 – 0.4. It is important to note that despite back to front disobedience sensitive nature it still remains the quickest at the assumed disobedience coefficient of 0.4. This also suggests the importance of airlines employing methods to increase obedience when disembarking as a 20% reduction in disobedience could cause up two minutes in extreme cases.

People Not Retrieving Bags	Reverse WMA	Back to Front	Random
0.2	411.9	374.7	468.2
0.4	352.7	325.3	352.7
0.6	289.9	261.0	309.8
0.8	234.1	215.7	235.6
1.0	200.1	200.9	200.0

Table 2: Sensitivity analysis of disembarking methods by changing people not retrieving bags

Table 2 shows a steady trend where the boarding time decreases and tends towards a constant time of 200s as the people not retrieving bags increases. This trend is important for two reasons. Firstly, it shows that if airlines could reduce the amount of bags carried it would result in much faster disembarking times, to the point it would no longer matter which disembarking method was employed. This is because less bags mean the aisle is blocked for a reduced amount of time. Even a minor increase in people not taking bags, for example from 40% to 60%, would result in a drastic reduction in disembarking time of 30s. This could be achieved by encouraging passengers to retrieve their bag in the period between when the plane lands and the disembarking process begins thus increasing the amount of people not retrieving during the disembarking.

4 Extension of Model to Other Aircraft

4.1 Flying Wing Aircraft

4.1.1 Flying Wing Boarding

The Flying Wing Aircraft has a revolutionary seating plan with an additional 3 aisles and 18 seats across, but only 14 rows. To account for this, we built upon the core algorithm of the narrow body in which passengers walk down the aisle, by simulating all 4 aisles at once, with an additional aisle connecting all of these at the top from the entrance. We initially considered simulating only one aisle and simply quartering the flow rate into the aisle. However, this is not realistic as the top aisle can still be blocked – for example, consider the case where a passenger is stowing their luggage in row 1 of the first aisle, whilst a passenger behind them waits to get into this aisle. Keeping with the assumption that only one passenger can fit into an aisle, such a scenario would block passengers from accessing all other aisles, increasing total boarding time. Thus, we must simulate all aisles boarding at once. Furthermore, although the number of aisles in this plane may cause confusion about where to go, we assumed that this would already be accounted for by the presence of flight attendants, causing no passengers to walk down the wrong aisle. The extended algorithm as experienced by a passenger is represented in the flow chart. A visualization of this model nearing completion is also displayed. Note: the top aisle is not included in this visualisation.

Different boarding models can be applied to the flying wing aircraft to different effect. Random and sectional boarding are relatively easily implemented, and both would be theoretically and practically effective. However, our optimal boarding method for the narrow body aircraft, the WMA method, is now rendered impractical to implement. When considering a seat block between two aisles, where ‘A’, ‘M’, and ‘W’ represent aisle, middle, and window seats respectively. Translating into rows six seats wide, you get the pattern A—M—W—W—M—A. It would be impractical for passengers to judge whether their seat is designated as A, M or W, even without incorporating groups.

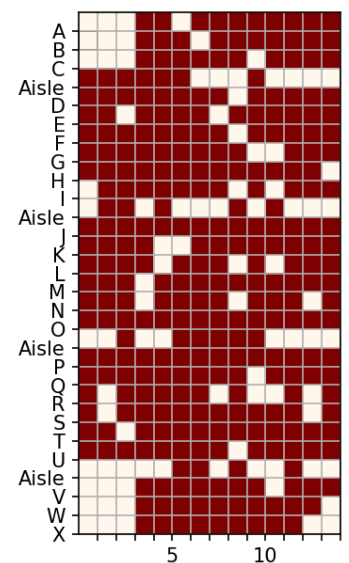


Figure 4.1: Flying Wing model

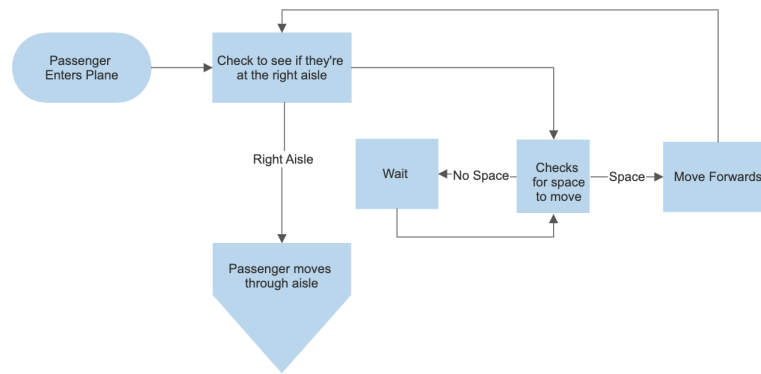


Figure 4.2: Flow diagram of passenger movement logic for the flying wing aircraft

Another potential boarding method we could adapt to a wide wing aircraft would be the modified Steffen method. However, given the established impracticality of the modified Steffen method on the narrow body, this would be even less realistic to expect passengers to follow it when the aircraft is boasting multiple aisles. Thus, we disregarded this method for this plane type. This left us with two viable boarding methods for the wing plane. These are shown in Fig 4.3, along with adjusted WMA times. The mean result times from these boarding methods were 593.2 seconds for the back to front time, 558.9 seconds for the random boarding, and 546.1 seconds for the adjusted WMA boarding time. Despite this, the optimal boarding method is the section boarding, from back to front. Despite having the lowest times, the impracticality of other solutions makes it the most attractive. The closest in terms of overall effectiveness would be random boarding. However, the organisational issues of trying to queue all the passengers in a random order with resulting in excessively large queues would more than out weigh the megre 34.3 second boarding time advantage.

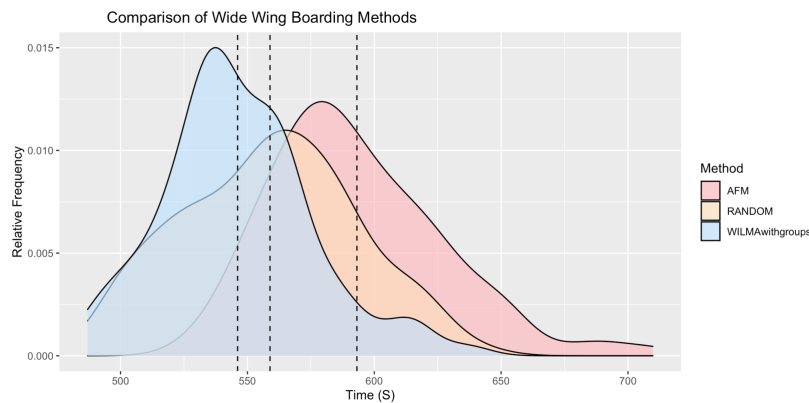
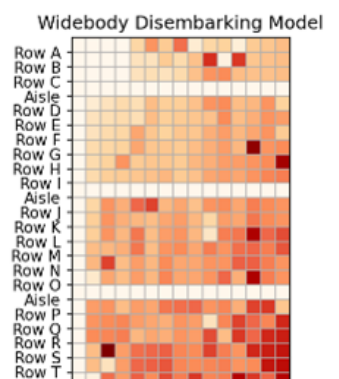


Figure 4.3: Comparison of distribution of times for different methods on the Flying Wing aircraft.

Note that the mean lines for AFM, random and WMA are located at 546.1, 558.9 and 593.2 respectively.

4.1.2 Flying Wing Disembarking

The core logic of the priority disembarking model remains the same when adapted to the Flying Wing Aircraft. The plane is broken into four sub-aisles that passengers are able to move into, dependent on the priority logic. Additionally a leaving aisle has been added that runs by all of the sub aisles. Passengers are moved into and along this leaving aisle to the exit through the use of the priority logic.



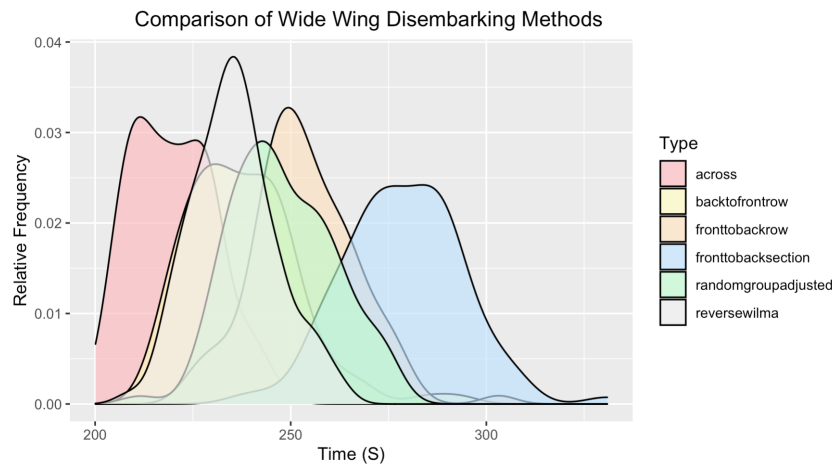


Figure 4.5: Comparison of disembarking methods on the flying wing aircraft

Similarly to the narrow body disembarking, models that prioritise aisle flow will be the fastest in the flying wing aircraft. The most successful model was the across method in Fig 4.4. The across method increased the priority assigned to passengers the further they are from the door, with respect to row and seat. This is in comparison to the ‘front to back’ disembarking method that only increases priority across the row. The across method of distributing priority resulted in the quickest disembarking times as it systematically emptied the aircraft from the bottom right to top left of the diagram (with the exit in the top left). This allowed most people to enter the aisle when they are few people behind them (as the people behind have already left). This means there are minimal aisle blockages when passengers retrieve bags, and passengers in front of the blockage will be able to move into the aisle. The across method is quite practical as it just requires for people to wait for the person behind to leave, or the person behind to retrieve a bag hence allowing them to leave. Similarly to other disembarking models, this is a form of ‘right of way system’, where people furthest from the door have right of way.

4.2 Two-Entrance Two-Aisle Aircraft

4.2.1 Two-Entrance Two-Aisle Boarding

The two-entrance two-aisle aircraft adds the complexity of multiple entrances, as well as a first-class section. However, we made the following assumptions:

- The first-class section would board first, as is standard across airlines. Any late passengers would interfere negligibly with the rest of the boarding process, as they do not walk down the same aisles as the rest of the passengers
- Rows 12 to 26 (and first-class) would board from the front entrance, whereas rows 27 to 47 would board from the back
- All passengers would board from the correct entrance. In a similar reasoning to everybody walking down the right aisle, we assumed that a plane of this size – and especially one with first class – would have sufficient flight attendants to ensure that this did not happen.

Given these assumptions, a valid simplification can be made to the model: the total boarding time would simply be the time taken to board first class, added to the greatest boarding time of the two sections (seats accessed from entrance 1 and seats accessed from entrance 2) of the plane. Boarding

methods in first class are unreasonable to implement in real life given the smaller number of seats, but also the easier access to seats due to greater space. Thus, this time was calculated using the random boarding method with increased speeds for walking and stowing luggage. The average was found to be We again tested this plane with unstructured (random) and sectional methods and our proposed method of Wilma with groups (in this case, the middle seat does not exist). Once again, Wilma with groups was found to be the most effective method.

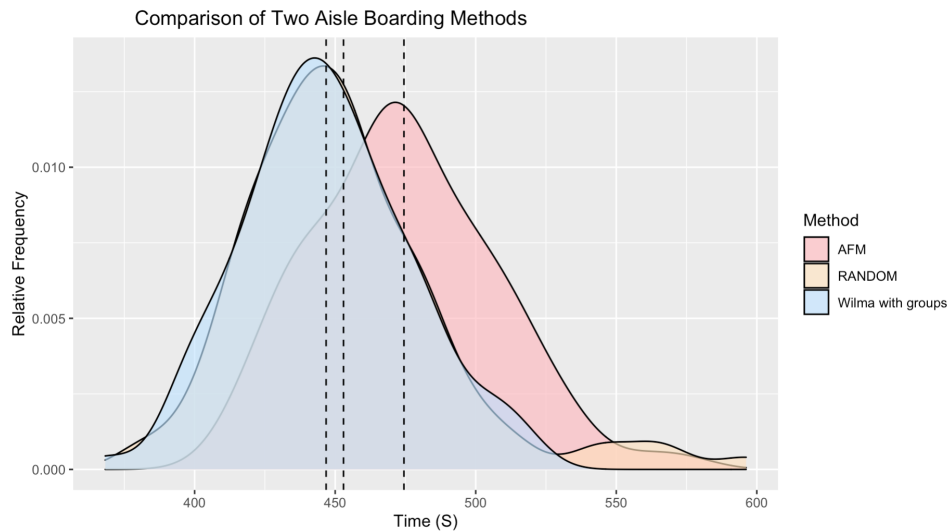


Figure 4.6: Comparison of boarding methods on the two-entrance two-aisle aircraft

4.2.2 Two-Entrance Two-Aisle Disembarking

The assumptions made in the boarding model of Two Entrance Two Aisle can be carried across to the disembarking model. Consequently, simulation were run on both halves of the plane for disembarking and the times were added to the time taken to board first class. Although disembarking occurs on two aisles in this model, aisle flow is still valued and therefore models that increase aisle flow such as back to front is still fastest with a mean time of 180s.

5 Pandemic Capacity Decrease

The COVID-19 pandemic has introduced many additional barriers to air travel. Notably, the passenger capacity of aircrafts is forced to decrease to help combat the spread of the disease. We test and present the effect on embarking and disembarking the three aircraft types when passenger capacity is limited to 70%, 50% and 30%. We ran 1000 test cases for all aircraft at all capacity levels with the three most optimal methods, the averages for which can be found in the appendix.

5.1 Boarding

An important consideration is that it is not simply random what tickets are not for sale on an aircraft with reduced capacity. Instead, they are chosen to maximise social distancing. For capacity c each row with numbers of seats s , we allowed a maximum of $\lceil c \times s \rceil$ seats to be filled in that row (where $\lceil x \rceil$ is the ceiling function), and reduced passengers randomly from there on until the number of passengers reached c . However, groups are still allowed to sit with each other. An analysis of the data would suggest that for both the Narrow Body and Flying Wing aircraft, the Wilma with groups method remains preferable up until a capacity of 50%. At this value its efficiency is only marginal. However, beyond this it becomes optimal to board by section (aft then middle then front). This

can be explained in practice, as at lower numbers of passengers, there is lesser chance of someone blocking the way to a seat – the main problem with sectional boarding. Therefore, without this problem, filling up from the back allows the most passengers to enter the queue at once, resulting in being more optimal. For the Two Entrance Two Aisle aircraft, sectional boarding also quickly becomes the most. This method has the added benefit of splitting passengers into boarding groups that also sit together, meaning that although contact cannot be completely avoided, it is minimized to be with the same people. If the pandemic is at the point that capacities of 50% or below need to be enforced, then this is a valuable aspect. Therefore, we would recommend this method for capacities of 50% and below on all aircraft. For capacities of 70% and 100%, the original recommended method remains the most optimal (this is still sectional for the Two Entrance Two Aisle aircraft).

5.2 Disembarking

To model reduced capacity of disembarking, we assumed a similar dispersal of passengers as in boarding. For all three aircraft, back to front remained the quickest method of disembarking no matter the capacity. Unfortunately, this does not preserve the social distancing between differing sections of the plane as achieved by boarding. However, at low capacities, disembarking times between different methods became exceedingly quick (under 2 minutes) and closer together. It would be no huge cost to the aircraft to favour a slower method at these capacities. Thus, at low capacities which aim to contain Covid, we would recommend a front to back method. Although this has not been modelled, by extrapolating data for modelled methods, it is clear that this would still be done in a tight timeframe.

6 Evaluation of Models

6.1 Strengths

- **Adaptable.** Not only can our models be used for many different boarding and disembarking methods, but our models can be adapted to wide range of plane shapes and sizes, that could consist of multiple aisles or entrances, with relative ease.
- **Comprehensive.** Our models take into account a wide range of factors affecting boarding and disembarking times, such as people moving past each other within rows, or time taken to stow and retrieve luggage. These times are calculated using real life data, ensuring the highest accuracy.
- **Realistic.** Many online models may have the strengths above, but fail to account for many common behaviours, such as people disobeying boarding instructions and passengers travelling together in groups.

6.2 Limitations

- **Memory intensive.** Due to boarding/disembarking being a stochastic process, a large number of test cases are needed to obtain an accurate average for any given scenario. Our model is bulky.
- **Large number of assumptions made.** For ease of simulation, we assumed many things, ranging from a constant walking speed down the aisle, or that passengers would always sit in the correct seat or walk down the correct aisle to their seat. In reality, this will not always be the case. To improve our model, we could include these factors in our simulation.

Letter to Executive

Dear Airline Executive,

Through analysis of boarding and disembarking methods, our team has developed a model that has allowed us to determine the optimal boarding and disembarking methods for a variety of different aircraft, with a number of different restrictions.

One consideration when deciding boarding methods is the impact that it will have on family groups. It is vital that we avoid splitting these groups when boarding to ensure all passengers have a positive travelling experience with you. Another factor that affects loading times is the number of passengers that don't follow boarding methods.

We found that three main factors that contribute to the time take to board an aircraft. The first of these is the walking speed of the passengers. However, there is not much that can realistically be done about this. Secondly, there is the time taken to stow overhead luggage. While passengers are doing this, they are blocking the aisle. Another aisle blockage comes passengers try to get to seats that are blocked by other passengers in same row.

These impact of these aisle blockages can be minimised by the chosen boarding method, and also by several different techniques. These include ensuring that people follow the boarding method (potentially through regulation from air stewards), and also by making more easily accessible overhead storage, to minimise time spent retrieving stowed luggage.

That being said, method choice is an easy way to immediately speed up passenger boarding/disembarking. In a standard narrow body aircraft, passengers should be boarded with the adjusted WMA method (window seats board first, followed by middle seats and then aisle seats, but groups board together), and should disembark giving the right of way to passengers coming from the back. Both methods minimise aisle blockages, and allow optimal aisle flow.

In a wide wing aircraft, the optimal boarding method is by section (back to front). This fastest method that can be practically implemented. To disembark, we recommend the 'across' method, similar the method for a narrow aircraft, where passengers furthest from the door get right of way. For the 2 aisle, 2 entrance aircraft, the recommended methods are the same as the narrow body: adjust WMA for boarding and back to front for disembarking. We hope these recommendations and explanations will aid you in running your airline, and look forward to your feedback.

7 Appendices

References

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YouTube videos used for analysis throughout the report

<https://www.youtube.com/watch?v=SnAYtU3nc0g>

<https://www.youtube.com/watch?v=F4Dt2-kc5Mw>

<https://www.youtube.com/watch?v=OCCUgcb4LvE>

<https://www.youtube.com/watch?v=V3RXT0D0CnM>

<https://www.youtube.com/watch?v=jv4y4ir07Nc>

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<https://www.youtube.com/watch?v=perToEokKR0>

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<https://www.youtube.com/watch?v=mrcQL5Od-Fg>

Appendix A

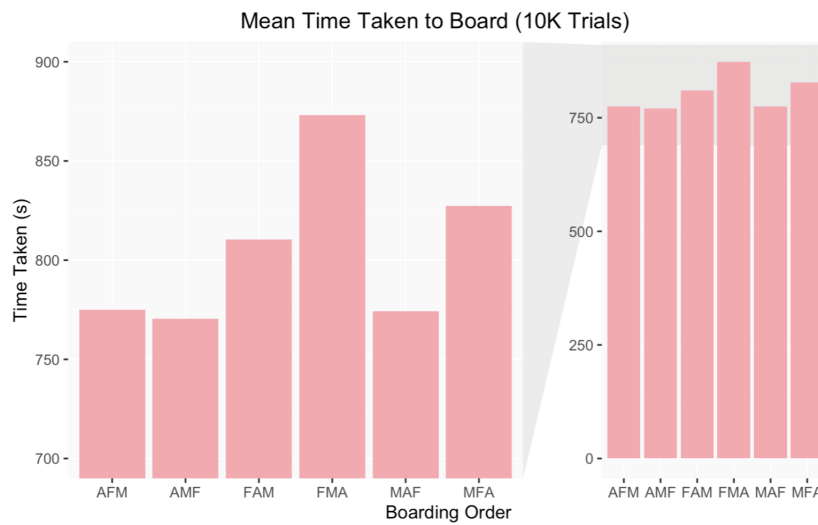


Figure 7.1: Bar chart of mean boarding times for different boarding by section combinations. The labels are in order of section boarding (e.g. AFM means aft first, front second, middle last. Note too that the chart on the left is an enlarged section of the chart on the right.

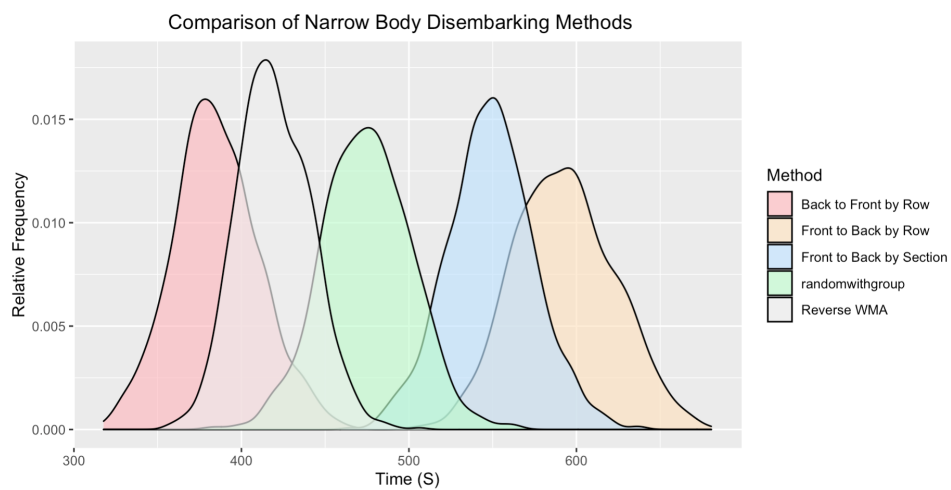


Figure 7.2: Comparison of disembarking methods for narrow body aircraft

Appendix B

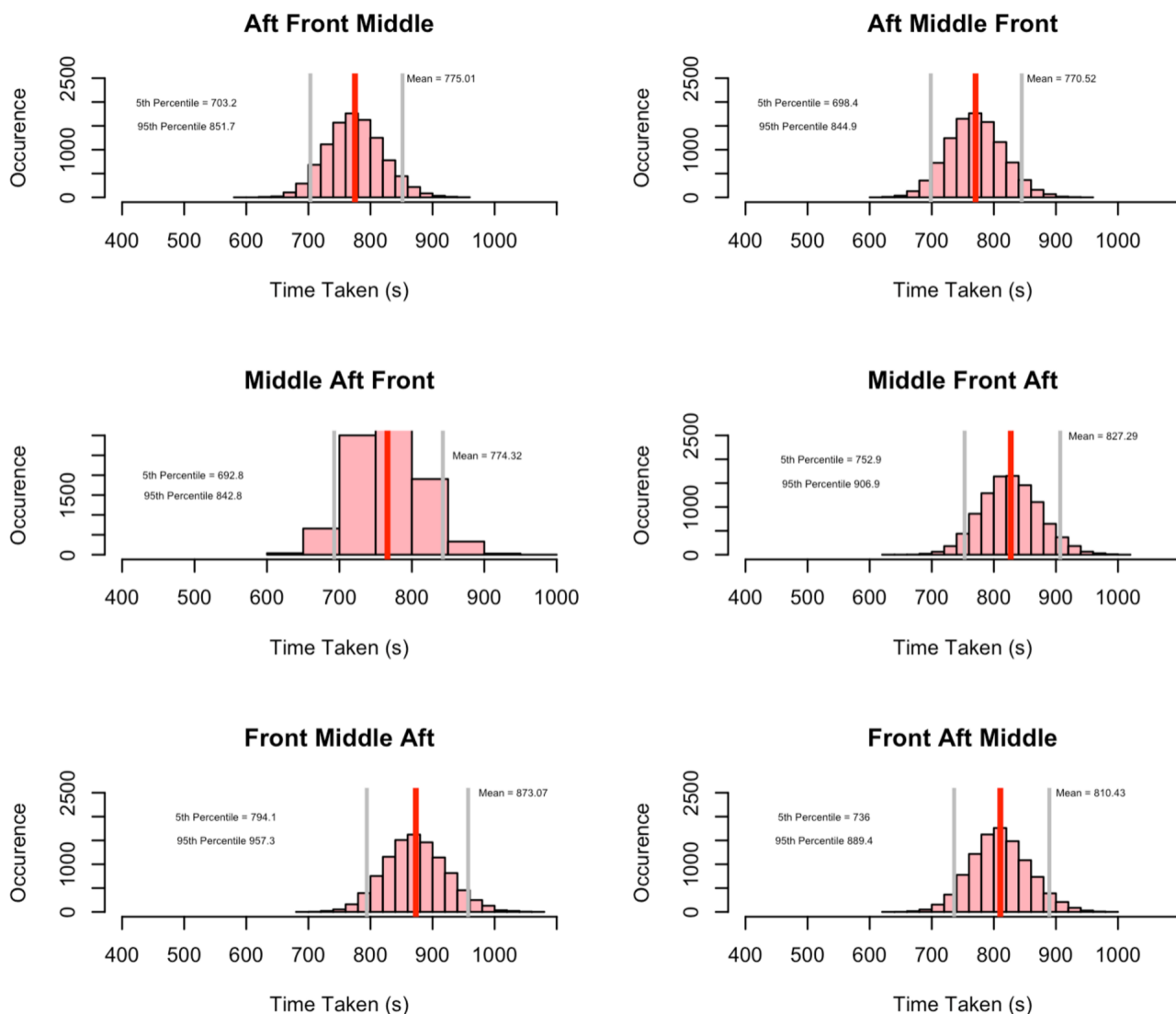


Figure 7.3: Monte Carlo simulation graphs of boarding by sections method in different orders. For instance, Aft Front Middle means first the aft is boarded, then front, and finally the middle.

Appendix C

			Narrow Body	Time taken depending on different boarding methods (s)
Capacity of plane	Wilma with groups	Random Boarding	Section boarding	
100%	644	700	764	
70%	545	550	600	
50%	301	311	302	
30%	194	202	180	

Flying Wing		Time taken depending on different boarding methods (s)	
Capacity of plane	Wilma with groups	Random Boarding	Section boarding
100%	493	552	575
70%	332	364	383
50%	237	263	254
30%	189	187	177

2E2A			
Capacity of plane	Wilma with groups	Random Boarding	Section boarding
100%	453	456	454
70%	319	302	292
50%	240	238	222
30%	179	177	152

COVID TALES DISEMBARKING

Narrow Body	Time taken depending on different boarding methods (s)		
Capacity of plane	Back to front	Reverse Wilma	Random
100%	384	419	474
70%	275	302	308
50%	128	224	220
30%	106	139	132

Flying Wing	Time taken depending on different boarding methods (s)		
Capacity of plane	Across (modified Back to Front)	Reverse Wilma	Random
100%	221	239	248
70%	167	160	178
50%	111	125	124
30%	95	97	106

2E2A	Time taken depending on different boarding methods (s)		
Capacity of plane	Back to front	Reverse Wilma	Random
100%	180	198	209
70%	142	168	192
50%	104	154	142
30%	87	92	104

Appendix D

Code used in R for statistical analysis

```

1 # INSTALL AND LOAD PACKAGES #####
2
3 library(datasets) # Load/unload base packages manually
4
5 ##### Graph Function #####
6 make_histogram <- function(x, colour, title, labelpos, meanpos, breaks1) {
7   quantile5 <- c(quantile(x, probs = 0.05))
8   quantile95 <- c(quantile(x, probs = 0.95))
9   mean <- mean(x)
10  hist(x,

```

```

11     xlim=c(400,1100),
12     ylim=c(0,350),
13     col=colour,
14     breaks = breaks1,
15     xlab="Time Taken (s)",
16     ylab="Occurence",
17     main=title)
18 # Add lines for mean, 5%, and 95%
19 abline(v = mean,
20        col = "red",
21        lwd = 3)
22 abline(v = c(quantile5, quantile95),
23        col = "grey",
24        lwd = 2)
25 # Add labels
26 text(x = mean*meanpos,
27      y = 125,
28      paste("Mean =", round(mean,2)),
29      col = "black",
30      cex = 1)
31 text(x = mean*labelpos,
32      y = 87.5,
33      paste("5th Percentile =", round(quantile5,1)),
34      col = "black",
35      cex = 0.7)
36 text(x = mean*labelpos,
37      y = 50,
38      paste("95th Percentile =", round(quantile95, 1)),
39      col = "black",
40      cex = 0.7)
41 }
42 dev.off()
43 #function(x,colour, title,labelpos,meanpos)
44 #####Finding best and worst boarding by sections #####
45 par(mfrow = c(3,2))
46 make_histogram(sectionsdata$afm, "#ffb3ba",
47                "Aft Front Middle",
48                0.65, 1.18 )
49 make_histogram(sectionsdata$amf, "#ffb3ba",
50                "Aft Middle Front",
51                0.65, 1.18 )
52 make_histogram(sectionsdata$maf, "#ffb3ba",
53                "Middle Aft Front",
54                0.65, 1.18 )
55 make_histogram(sectionsdata$mfa, "#ffb3ba",
56                "Middle Front Aft",
57                0.65, 1.18 )
58 make_histogram(sectionsdata$fma, "#ffb3ba",
59                "Front Middle Aft",
60                0.65, 1.18 )
61 make_histogram(sectionsdata$fam, "#ffb3ba",
62                "Front Aft Middle",
63                0.65, 1.18 )
64 fma<-mean(sectionsdata$fma)
65 #Compare with groups to without gorups
66 dev.off()
67 #function(x,colour, title,labelpos,meanpos)
68 #Compare four boarding methods: Random, Seats, Best Section, Worst Section
69 par(mfrow = c(1,2))
70 breaks2<-rep(10,1)

```

```

71 make_histogram(dataogmethod$Random, "#baffc9",
72               "Random Boarding",
73               1.33, 1.33 )
74 make_histogram(dataogmethod$Seat, "#bae1ff",
75               "Boarding by Seat: Groups",
76               1.45, 1.45, c
77               (460,485,510,535,560,585,610,635,660,685,710,735,760,785,810) )
78 make_histogram(dataogmethod$fma, "#ffb3ba",
79               "Front to Back",
80               0.65, 0.65 )
81 make_histogram(dataogmethod$amf, "#ffb3ba",
82               "Back to Front",
83               0.7, 0.7 )
84 make_histogram(dataogmethod$Seats.No.Groups, "#bae1ff",
85               "Boarding by Seat: No Groups",
86               1.6, 1.6, c(420,440,460,480,500,520,540,560,580,600))
87 dev.off()
88 make_histogram(groupsfirst$prioritize_groups_boarding, "#ffffba",
89               "Groups First Boarding",
90               1.28, 1.28 )
91 make_histogram(modifiedsteffen$modsteffen, "#ffdfba",
92               "Steffen Modified Boarding",
93               1.3, 1.3 )
94 #####Overlay histograms#####
95 library(ggplot2)
96 library(plyr)
97 #mean
98 mu <- ddply(dataogmethod.transp, "Type", summarise, grp.mean=mean(Time))
99 # Basic density
100 #ggplot(dataogmethod.transp, aes(x=Time, fill=Type)) +
101 #   geom_density(color="darkblue", fill="lightblue")
102 # Use semi-transparent fill
103 p <- ggplot(dataogmethod.transp, aes(x = Time, fill = Type)) +
104   geom_density(alpha=0.6) +
105   #theme_bw() +
106   #geom_vline(data=mu, aes(xintercept=grp.mean, color="black"),
107               #linetype="dashed") +
108   labs(title="
109           Comparison of Methods",
110         x="Time (S)", y = "Relative Frequency")
111 # Add mean lines
112 p+scale_fill_manual(values=c("#ffb3ba", "#ffffba","#ffdfba","#bae1ff","#baffc9"))
113 p
114 #####Provided Methods Sensitivity Analysis#####
115 dev.off()
116 attach(bagsensitivity)
117 linearModel <- lm(SectionBTF ~ NBC, data=sensitivityanalysis)
118 summary(linearModel)
119 lm(formula = Seats.No.Group ~ NBC)
120 sensitivityanalysis$NBC2 <- sensitivityanalysis$NBC^2
121 quadraticModel <- lm(SectionBTF ~ NBC + NBC2, data=sensitivityanalysis)
122 summary(quadraticModel)
123 #create sequence of hour values
124 NBCValues <- seq(0, 1, 0.01)
125 #create list of predicted time values using quadratic model
126 timePredict <- predict(quadraticModel, list(NBC=NBCValues, NBC2=NBCValues^2))
127 plot(bagsensitivity$bagcoef, bagsensitivity$section, pch = 19, cex = 0.75,
128      col = "#ffb3ba", xlab = "Bag Coefficient",
129      ylab = "Time (s)",
130      ylim=c(450,900),

```

```

130     main = "Sensitivity Analysis of 3 Given Models")
131 #lines(NBCValues, timePredict, col='#ffb3ba')
132 abline(lm(bagsensitivity$section ~ bagsensitivity$bagcoef), col = "#ffb3ba")
133 points(bagsensitivity$bagcoef, bagsensitivity$random, pch = 19, cex = 0.75, col =
134         "#95cca1")
135 abline(lm(bagsensitivity$random ~ bagsensitivity$bagcoef), col = "#95cca1")
136 #lines(bagsensitivity$bagcoef, bagsensitivity$wma, col="#bae1ff")
137 points(bagsensitivity$bagcoef, bagsensitivity$wma, pch = 19, cex = 0.75, col = "#
138         bae1ff")
139 abline(lm(bagsensitivity$wma ~ bagsensitivity$bagcoef), col = "#bae1ff")
140 legend(0.7, 550, legend=c("Back to Front", "Random", "Seat (No Groups)"), col=c("
141         #ffb3ba", "#95cca1", "#bae1ff"), lty=1:2, cex=0.65)
142 #####Suggested Methods Sensitivity#####
143 dev.off()
144 attach(seat_with_groups1)
145 plot(NBC, SeatwithGroups, pch = 19, cex = 0.75,
146       col = "#bae1ff", xlab = "Disobedience Coefficient",
147       ylab = "Time (s)",
148       ylim=c(550,750),
149       main = "Sensitivity Analysis of Chosen Method Against Random")
150 abline(lm(SeatwithGroups ~ NBC), col = "#bae1ff")
151 lines(NBC, Random, col="#95cca1")
152 points(NBC, Random, pch = 19, cex = 0.75, col = "#95cca1")
153 legend(0.75, 625, legend=c("Random", "Seat (Groups)"), col=c("#95cca1", "#bae1ff"
154         ), lty=1, cex=0.65)
155 #####
156 barplot((Barplot_section_order$Mean),
157         main="Mean Time Taken to Board (10k Trials)",
158         names.arg=c("AFM", "AMF", "MAF", "MFA", "FMA", "FAM"),
159         ylim=c(700,900),
160         xpd=FALSE,
161         col="#ffb3ba",
162         ylab = "Time Taken to Board (s)",
163         xlab = "Boarding Method")
164 #####
165 install.packages("ggforce")
166 library("ggforce")
167 attach(Barplot_section_order)
168 ggplot(Barplot_section_order, aes(Order, Mean)) +
169       # ggplot2 facet_
170       zoom plot
171       geom_bar(stat = "identity", fill = "#f2aab1") +
172       ggtitle("Mean Time Taken to Board (10K Trials)") +
173       labs(y= "Time Taken (s)", x = "Boarding Order") +
174       theme(panel.background = element_rect("#f9f9f9")) +
175       facet_zoom(ylim = c(700, 900), show.area=TRUE)
176 #####

```

Appendix E

```

1
2 from audioop import reverse
3 import graphlib
4 from lib2to3.pgen2.token import NUMBER
5 import random
6 from string import ascii_letters
7 import math
8

```

```
9 #Comment out if not using
10 import matplotlib.pyplot as plt
11 import numpy
12 import matplotlib.colors as colors
13 import matplotlib as mpl
14 from matplotlib.colors import Colormap, LinearSegmentedColormap, ListedColormap
15 #Constants for references
16 PRIORITY = 0
17 INTERNAL_COCK = 1
18 HAS_LUGGAGE = 2
19
20 # all measured in standard units (m,s,m/s etc)
21 AVERAGE_WALKING_SPEED = 0.8
22 AVERAGE_SEAT_PITCH = 0.78
23 TIME_TO_MOVE = AVERAGE_SEAT_PITCH / AVERAGE_WALKING_SPEED
24 TIME_TO_MOVE_PAST_SEAT = 2
25
26 #Priority system
27 priority_weightings = []
28 highest_priority_assigned = 0
29 #Things to change
30 BAG_COEFFICIENT = (20,80,10)
31 NAUGHTY_BOY_COEFFICIENT = 0.3
32 THANOS_SNAP_COEFFICIENT = 0.5
33 # proportions of group sizes
34 SINGLE_PRINGLE_COEFFICIENT = 70
35 COUPLES_COEFFICIENT = 20
36 THREESOME_COEFFICIENT = 10
37
38
39
40 # General setup shotput all seats in
41 NUM_ROWS = 33
42 NUM_SEATS = 6
43 AISLE_INDEX = 3
44 #Wide body shot
45 WIDE_WING_SEATS = 28
46 WIDE_WING_ROWS = 15
47 TWO_SEATS = 9
48 TWO_ROWS = 42
49 TWO_A_ROWS = 18
50 TWO_B_ROWS = 21
51 GAP_SIZE = 3
52
53
54
55
56 #Normal render
57 def intalize_render():
58
59     global highest_priority_assigned
60     #Absolute mess of code
61     image = []
62     for i in range(NUM_SEATS+1):
63         subimage = []
64         for k in range(NUM_ROWS):
65             if k % 2 == 0:
66                 subimage.append(0.5)
67             else:
68                 subimage.append(1.5)
```



```
69     image.append(subimage)
70
71
72
73     fig,ax = plt.subplots(1,1)
74
75     cmap = mpl.cm.OrRd
76     norm = mpl.colors.Normalize(vmin=-1, vmax=highest_priority_assigned)
77
78
79
80     image = numpy.array(image)
81     im = ax.imshow(image, cmap=cmap, norm = norm)
82
83     ax.set_yticks(numpy.arange(0.5, NUM_SEATS+1.5, 1).tolist(), minor=False)
84     ax.yaxis.grid(True, which='major')
85     ax.set_yticklabels(['Row A', 'Row B', 'Row C', 'Aisle', 'Row D', 'Row E', 'Row F'])
86     ax.set_ylim(top=-0.5)
87
88     ax.set_xticks(numpy.arange(0.5, NUM_ROWS+1.5, 1).tolist(), minor=False)
89     ax.xaxis.grid(True, which='major')
90     xticklist = []
91     #Create list of numbers between
92     for i in range(NUM_ROWS):
93         if ((i+1) % 5 == 0) and (i != 0):
94             xticklist.append(str(i+1))
95         else:
96             xticklist.append('')
97
98     ax.set_xticklabels(xticklist)
99     ax.set_xlim(left=-0.5)
100
101     return im,fig
102 def update_render(seat_plan):
103
104     visualizer = []
105
106     for i,column in enumerate(seat_plan):
107         visualizer.append([])
108         for seats in column:
109
110             visualizer[i].append(seats[PRIORITY])
111
112
113
114
115
116     im.set_data(visualizer)
117     fig.canvas.draw_idle()
118     plt.pause(1)
119 def intalize_render_two_thing():
120
121     global highest_priority_assigned
122
123     #Absolute mess of code
124     image = []
125     for i in range(TWO_SEATS):
126         subimage = []
127         for k in range(TWO_ROWS):
128             if k % 2 == 0:
```

```
129         subimage.append(0.5)
130     else:
131         subimage.append(1.5)
132
133     image.append(subimage)
134
135
136 fig,ax = plt.subplots(1,1)
137
138 cmap = mpl.cm.OrRd
139 norm = mpl.colors.Normalize(vmin=-1, vmax=highest_priority_assigned)
140
141
142
143 image = numpy.array(image)
144 im = ax.imshow(image, cmap=cmap, norm = norm)
145
146 ax.set_yticks(numpy.arange(0.5, TWO_SEATS+0.5, 1).tolist(), minor=False)
147 ax.yaxis.grid(True, which='major')
148 ax.set_yticklabels(['Row A', 'Row B', 'Aisle', 'Row C', 'Row D', 'Row E', 'Aisle', '
149 Row F', 'Row G'])
150 ax.set_ylim(top=-0.5)
151 ax.set_title('Two Doors Two Aisles Disembarking Model')
152 ax.set_xticks(numpy.arange(0.5, TWO_ROWS+.5, 1).tolist(), minor=False)
153 ax.xaxis.grid(True, which='major')
154 xticklist = []
155 #Create list of numbers between
156 for i in range(TWO_ROWS):
157     if ((i+1) % 5 == 0) and (i != 0):
158         xticklist.append(str(i+1))
159     else:
160         xticklist.append('')
161
162 ax.set_xticklabels(xticklist)
163 ax.set_xlim(left=-0.5)
164
165 return im,fig
166
167 def intalize_render_widebody():
168
169     global highest_priority_assigned
170
171     #Absolute mess of code
172     image = []
173     for i in range(WIDE_WING_SEATS):
174         subimage = []
175         for k in range(WIDE_WING_ROWS):
176             if k % 2 == 0:
177                 subimage.append(0.5)
178             else:
179                 subimage.append(1.5)
180
181         image.append(subimage)
182
183
184 fig,ax = plt.subplots(1,1)
185
186 cmap = mpl.cm.OrRd
187 norm = mpl.colors.Normalize(vmin=-1, vmax=highest_priority_assigned)
```

```

188
189 image = numpy.array(image)
190 im = ax.imshow(image, cmap=cmap, norm = norm)
191
192 ax.set_yticks(numpy.arange(0.5, WIDE_WING_SEATS+0.5, 1).tolist(), minor=False
)
193 ax.yaxis.grid(True, which='major')
194 ax.set_yticklabels(['Row A', 'Row B', 'Row C', 'Aisle', 'Row D', 'Row E', 'Row F', '
Row G', 'Row H', 'Row I', 'Aisle', 'Row J', 'Row K', 'Row L', 'Row M', 'Row N', 'Row O'
, 'Aisle', 'Row P', 'Row Q', 'Row R', 'Row S', 'Row T', 'Row U', 'Aisle', 'Row V', 'Row
W', 'Row X'])
195 ax.set_ylim(top=-0.5)
196 ax.set_title('Widebody Disembarking Model')
197 ax.set_xticks(numpy.arange(0.5, WIDE_WING_ROWS+0.5, 1).tolist(), minor=False)
198 ax.xaxis.grid(True, which='major')
199 xticklist = []
200 #Create list of numbers between
201 for i in range(WIDE_WING_ROWS):
202     if ((i+1) % 5 == 0) and (i != 0):
203         xticklist.append(str(i+1))
204     else:
205         xticklist.append('')
206
207 ax.set_xticklabels(xticklist)
208 ax.set_xlim(left=-0.5)
209
210 return im,fig
211
212 #General shot to setup
213 def generate_priorities(highest_priority_assigned):
214
215
216     weights = list(range(1, highest_priority_assigned+1))
217
218     return(weights)
219 def group_size():
220     return random.choices([1,2,3], weights=(SINGLE_PRINGLE_COEFFICIENT,
COUPLES_COEFFICIENT, THREESOME_COEFFICIENT), k=1)[0]
221 def assign_luggage():
222     return random.choices([0,1,2], weights=BAG_COEFFICIENT, k=1)[0]
223 def bag_shit():
224     global seating_plan
225     #Intalize bag amounts
226     lockers = [[0,0] for i in range(NUM_ROWS)]
227     for row in range(NUM_ROWS):
228         for seat in range(NUM_SEATS+1):
229
230             if seat < 3:
231                 lockers[row][0] += seating_plan[seat][row][HAS_LUGGAGE]
232
233             elif seat > 3:
234                 lockers[row][1] += seating_plan[seat][row][HAS_LUGGAGE]
235     return lockers
236 #Widebody
237 def bag_shit_wide():
238     global seating_plan
239     #Intalize bag amounts
240
241     lockers = [[ [0,0] for _ in range(WIDE_WING_ROWS-1)] for _ in range(4)]
242

```

```

243
244     for row in range(1,WIDE_WING_ROWS):
245
246         for seat in range(1,WIDE_WING_SEATS):
247             sublocker = math.floor((seat)/7)
248             if seating_plan[seat][row][PRIORITY] != -1:
249                 if seat % 7 < 3:
250                     lockers[sublocker][row-1][0] += seating_plan[seat][row][
HAS_LUGGAGE]
251
252                 elif seat % 7 > 3:
253                     lockers[sublocker][row-1][1] += seating_plan[seat][row][
HAS_LUGGAGE]
254
255
256
257
258
259
260     return lockers
261 #Modifying shot
262 def group_shit():
263     for row in range(NUM_ROWS):
264         #Resets var
265         current_group_size = 0
266         current_group_priority = []
267         current_group_people_added = 0
268         for seat in range(NUM_SEATS+1):
269             #make sure we not in aisles
270             if seat != AISLE_INDEX:
271                 #If not currently generating create a new group
272                 if current_group_people_added == 0:
273                     current_group_size = current_group_people_added =
group_size()
274
275                     if current_group_people_added == 1:
276                         current_group_people_added = 0
277                     else:
278                         current_group_priority.append(seating_plan[seat][row][
PRIORITY])
279
280                         current_group_people_added -=1
281
282                 else: # Currently generating a group
283                     current_group_priority.append(seating_plan[seat][row][
PRIORITY])
284
285                     current_group_people_added -=1
286
287                 #If all people added to group
288                 if current_group_people_added == 0:
289                     #Loop back through people and send priority to
average
290
291                     gone_through_aisles = 0
292                     for i in range(current_group_size):
293                         #Go back through and adjust priority
294
295                         #If gone through aisles add another
296                         if (seat - i) == AISLE_INDEX:
297                             gone_through_aisles = 1

```

```

297         seating_plan[seat-(i+gone_through_aisles)][row][
298     PRIORITY] = round(sum(current_group_priority) / len(current_group_priority))
299 def naughty_people():
300
301     #Generate total amount of naughty boys
302     naughty_bois = math.ceil(NUM_ROWS*NUM_SEATS*NAUGHTY_BOY_COEFFICIENT)
303
304     for i in range(naughty_bois):
305         numbers = list(range(0, NUM_SEATS+1))
306
307         numbers.remove(3)
308
309         seat = random.choice(numbers)
310         row = random.randrange(NUM_ROWS)
311
312         seating_plan[seat][row][PRIORITY] = random.randrange(1,
highest_priority_assigned+1)
313 def thanos_snap():
314     for seat in range(NUM_SEATS+1):
315         for row in range(NUM_ROWS):
316             if THANOS_SNAP_COEFFICIENT > random.random():
317                 seating_plan[seat][row] = [-1,0]
318 #Narrow body boarding
319 def reverse_wilma():
320     global seating_plan
321     global highest_priority_assigned
322
323     highest_priority_assigned = 3
324
325     seating_plan = [[ [3,0,assign_luggage()] for _ in range(NUM_ROWS)] for _ in
range(NUM_SEATS + 1)]
326     seating_plan[0]= [[1,0,assign_luggage()] for _ in range(NUM_ROWS)]
327     seating_plan[6]= [[1,0,assign_luggage()] for _ in range(NUM_ROWS)]
328     seating_plan[1]= [[2,0,assign_luggage()] for _ in range(NUM_ROWS)]
329     seating_plan[5]= [[2,0,assign_luggage()] for _ in range(NUM_ROWS)]
330     seating_plan[AISLE_INDEX]= [[-1,0] for _ in range(NUM_ROWS)]
331
332     naughty_people()
333     group_shit()
334 def random_deboard():
335     global seating_plan
336     global highest_priority_assigned
337
338     highest_priority_assigned = 10
339
340     seating_plan = [[ [random.randrange(1,10),0,assign_luggage()] for _ in range(
NUM_ROWS)] for _ in range(NUM_SEATS + 1)]
341     seating_plan[AISLE_INDEX]= [[-1,0] for _ in range(NUM_ROWS)]
342
343     group_shit()
344 def sections():
345
346     global seating_plan
347
348     global highest_priority_assigned
349
350     highest_priority_assigned = 3
351     fjuk = []
352     for i in range(NUM_SEATS+1):

```

```
353     aisles = []
354     for k in range(0,11):
355         aisles.append([3,0,assign_luggage()])
356     for k in range(11,22):
357         aisles.append([2,0,assign_luggage()])
358     for k in range(22,NUM_ROWS):
359         aisles.append([1,0,assign_luggage()])
360     fjuk.append(aisles)
361     seating_plan = fjuk
362     seating_plan[AISLE_INDEX]= [[-1,0] for _ in range(NUM_ROWS)]
363     naughty_people()
364     group_shit()
365 def back_to_front():
366     global seating_plan
367     global highest_priority_assigned
368
369
370     #Create empty seating plan
371     seating_plan = [[ [-1,0,assign_luggage()] for _ in range(NUM_ROWS)] for _ in
range(NUM_SEATS + 1)]
372
373     highest_priority_assigned = 0
374
375     for row in range(NUM_ROWS):
376         for seat in range(NUM_SEATS+1):
377             #Increment priority
378
379
380             if seat == 0 or seat == 3:
381                 highest_priority_assigned += 1
382                 seating_plan[seat][row] = ([highest_priority_assigned,0,
assign_luggage()])
383
384     seating_plan[AISLE_INDEX]= [[-1,0] for _ in range(NUM_ROWS)]
385     #naughty_people()
386     #group_shit()
387 def generate_front_to_back():
388     global seating_plan
389     global highest_priority_assigned
390
391
392     #Create empty seating plan
393     seating_plan = [[ [-1,0,assign_luggage()] for _ in range(NUM_ROWS)] for _ in
range(NUM_SEATS + 1)]
394
395     highest_priority_assigned = 0
396
397     for row in reversed(range(NUM_ROWS)):
398         for seat in range(NUM_SEATS+1):
399             #Increment priority
400
401
402             if seat == 0 or seat == 3:
403                 highest_priority_assigned += 1
404                 seating_plan[seat][row] = ([highest_priority_assigned,0,
assign_luggage()])
405
406     seating_plan[AISLE_INDEX]= [[-1,0] for _ in range(NUM_ROWS)]
407
408
```

```

409     naughty_people()
410     group_shit()
411
412 def group_shit_two():
413     for row in range(4, TWO_ROWS):
414         #Resets var
415         current_group_size = 0
416         current_group_priority = []
417         current_group_people_added = 0
418
419
420
421         for seat in range(0,9):
422
423             #make sure we not in aisles
424             if (seat != 2 or seat !=4) and seating_plan[seat][row][PRIORITY] !=
-1:
425                 #If not currently generating create a new group
426                 if current_group_people_added == 0:
427                     current_group_size = current_group_people_added = group_size
428                 ()
429                 if current_group_people_added == 1:
430                     current_group_people_added = 0
431                 else:
432                     current_group_priority.append(seating_plan[seat][row][
PRIORITY])
433                     current_group_people_added -=1
434
435                 else: # Currently generating a group
436                     current_group_priority.append(seating_plan[seat][row][PRIORITY
])
437                     current_group_people_added -=1
438
439                 #If all people added to group
440                 if current_group_people_added == 0:
441                     #Loop back through people and send priority to average
442
443
444                     gone_through_aisles = 0
445                     for i in range(current_group_size):
446                         #Go back through and adjust priority
447
448                         #If gone through aisles add another
449                         if (seat - i) == 2 or (seat - i) == 6:
450                             gone_through_aisles = 1
451
452                             seating_plan[(seat-(i+gone_through_aisles))][row][
PRIORITY] = round(sum(current_group_priority) / len(current_group_priority))
453 #Wide body boarding
454 def clear_aisles_widebody():
455     global seating_plan
456
457
458     #Clear aisles
459     for i in range(WIDE_WING_SEATS):
460         if i % 7 == 3 and i != WIDE_WING_SEATS:
461             seating_plan[i]= [[-1,0] for _ in range(WIDE_WING_ROWS)]
462     #Clear front aisles
463     for k in range(WIDE_WING_SEATS):

```



```

518                                     #Loop back through people and send priority to
average
519
520
521                                     gone_through_aisles = 0
522                                     for i in range(current_group_size):
523                                         #Go back through and adjust priority
524
525                                         #If gone through aisles add another
526                                         if (seat - i) == 3:
527                                             gone_through_aisles = 1
528
529                                     seating_plan[(seat-(i+gone_through_aisles))+
current_aisle][row][PRIORITY] = round(sum(current_group_priority) / len(
current_group_priority))
530 def reverse_wilma_widebody():
531     global seating_plan
532     global highest_priority_assigned
533
534     highest_priority_assigned = 3
535
536     #Make an empty
537     seating_plan = [[ [-1,0,0] for _ in range(WIDE_WING_ROWS)] for _ in range(
WIDE_WING_SEATS)]
538
539     for i in range(WIDE_WING_SEATS):
540         if i in [2,4,9,11,16,18,23,25]:
541             seating_plan[i]= [[3,0,assign_luggage()] for _ in range(
WIDE_WING_ROWS)]
542         elif i in [1,5,8,12,15,19,22,26]:
543             seating_plan[i]= [[2,0,assign_luggage()] for _ in range(
WIDE_WING_ROWS)]
544         elif i in [0,6,7,13,14,20,21,27]:
545             seating_plan[i]= [[1,0,assign_luggage()] for _ in range(
WIDE_WING_ROWS)]
546
547
548     clear_aisles_widebody()
549     naughty_people_wide()
550     group_shit_wide()
551 def random_deboard_widebody():
552     global seating_plan
553     global highest_priority_assigned
554
555     highest_priority_assigned = 10
556
557     seating_plan = [[ [ random.randrange(1,10),0,assign_luggage()] for _ in range
(WIDE_WING_ROWS)] for _ in range(WIDE_WING_SEATS)]
558
559     clear_aisles_widebody()
560     group_shit_wide()
561 def sections_widebody():
562
563     global seating_plan
564
565     global highest_priority_assigned
566
567     highest_priority_assigned = 3
568     fjuk = []
569     for i in range(WIDE_WING_SEATS):

```

```

570     aisles = []
571     for k in range(1,8):
572         aisles.append([3,0,assign_luggage()])
573     for k in range(8,12):
574         aisles.append([2,0,assign_luggage()])
575     for k in range(12,WIDE_WING_ROWS+1):
576         aisles.append([1,0,assign_luggage()])
577     fjuk.append(aisles)
578     seating_plan = fjuk
579
580
581
582     clear_aisles_widebody()
583
584     naughty_people_wide()
585     group_shit_wide()
586 def back_to_front_widebody():
587     global seating_plan
588     global highest_priority_assigned
589
590
591     #Create empty seating plan
592     seating_plan = [[ [k+j,0,assign_luggage()] for k in range(WIDE_WING_ROWS)]
593     for j in range(WIDE_WING_SEATS)]
594
595     highest_priority_assigned = WIDE_WING_SEATS+WIDE_WING_ROWS
596
597     clear_aisles_widebody()
598
599     naughty_people_wide()
600     group_shit_wide()
601 def across_widebody():
602     global seating_plan
603     global highest_priority_assigned
604
605     #Create empty seating plan
606     seating_plan = [[ [k,0,assign_luggage()] for k in range(WIDE_WING_ROWS)] for
607     j in range(WIDE_WING_SEATS)]
608
609     highest_priority_assigned = WIDE_WING_ROWS
610
611     clear_aisles_widebody()
612
613     naughty_people_wide()
614     group_shit_wide()
615 def naughty_people_two ():
616     global seating_plan
617     global highest_priority_assigned
618
619     #Generate total amount of naughty boys
620     naughty_bois = math.ceil(((TWO_SEATS-2)*(TWO_A_ROWS+TWO_B_ROWS)+18)*
621     NAUGHTY_BOY_COEFFICIENT)
622
623     for i in range(naughty_bois):
624
625         while True:
626             seat = random.randrange(TWO_SEATS)
627             row = random.randrange(TWO_ROWS)

```

```

627         if seating_plan[seat][row][PRIORITY] != -1:
628             seating_plan[seat][row][PRIORITY] = random.randrange(1,
629 highest_priority_assigned)
630         break
631 def generate_front_to_back_widebody ():
632     global seating_plan
633     global highest_priority_assigned
634
635     highest_priority_assigned = WIDE_WING_SEATS+WIDE_WING_ROWS
636     #Create empty seating plan
637     seating_plan = [[ [highest_priority_assigned-(k+j),0,assign_luggage()] for k
in range(WIDE_WING_ROWS)] for j in range(WIDE_WING_SEATS)]
638
639
640
641     clear_aisles_widebody()
642
643     naughty_people_wide()
644     group_shit_wide()
645
646
647 def bag_shit_Two():
648     global seating_plan
649     #Intalize bag amounts
650     lockers = [[0,0] for _ in range(TWO_A_ROWS)],[[0,0] for _ in range(
TWO_B_ROWS)]]
651
652     #A first
653     for row in range(TWO_A_ROWS):
654
655         for seat in range(TWO_SEATS):
656
657             if seating_plan[seat][row][PRIORITY] != -1:
658                 if seat <=3:
659                     lockers[0][row][0] += seating_plan[seat][row][HAS_LUGGAGE]
660
661                 elif seat > 3:
662                     lockers[0][row][1] += seating_plan[seat][row][HAS_LUGGAGE]
663     #B second
664     for row in range(TWO_B_ROWS):
665
666         for seat in range(TWO_SEATS):
667
668             if seating_plan[seat][row+GAP_SIZE+TWO_A_ROWS][PRIORITY] != -1:
669                 if seat <=3:
670                     lockers[1][row][0] += seating_plan[seat][row+GAP_SIZE+
TWO_A_ROWS][HAS_LUGGAGE]
671
672                 elif seat > 3:
673                     lockers[1][row][1] += seating_plan[seat][row+GAP_SIZE+
TWO_A_ROWS][HAS_LUGGAGE]
674
675
676
677
678
679
680     return lockers
681 def two_first_class():

```

```
682 global seating_plan
683 global highest_priority_assigned
684
685 #Generate first class
686 seating_plan[0][0][0] = highest_priority_assigned
687 seating_plan[0][1][0] = highest_priority_assigned
688 seating_plan[0][2][0] = highest_priority_assigned
689 seating_plan[1][0][0] = highest_priority_assigned
690 seating_plan[1][1][0] = highest_priority_assigned
691 seating_plan[1][2][0] = highest_priority_assigned
692 seating_plan[3][0][0] = highest_priority_assigned
693 seating_plan[3][1][0] = highest_priority_assigned
694 seating_plan[3][2][0] = highest_priority_assigned
695 seating_plan[5][0][0] = highest_priority_assigned
696 seating_plan[5][1][0] = highest_priority_assigned
697 seating_plan[5][2][0] = highest_priority_assigned
698 seating_plan[7][0][0] = highest_priority_assigned
699 seating_plan[7][1][0] = highest_priority_assigned
700 seating_plan[7][2][0] = highest_priority_assigned
701 seating_plan[8][0][0] = highest_priority_assigned
702 seating_plan[8][1][0] = highest_priority_assigned
703 seating_plan[8][2][0] = highest_priority_assigned
704 #Two
705 def two_cleanup():
706     global seating_plan
707     global highest_priority_assigned
708
709
710
711 #Clear aisles
712 seating_plan[2]= [[-1,0] for _ in range(TWO_ROWS)]
713 seating_plan[6]= [[-1,0] for _ in range(TWO_ROWS)]
714 #Tidy up first class
715 seating_plan[4][0] = [-1,0]
716 seating_plan[4][1] = [-1,0]
717 seating_plan[4][2] = [-1,0]
718
719
720 #Clear queues out
721 for k in range(TWO_SEATS):
722     for j in range(TWO_ROWS):
723         if j in [3, 18,19,20,42]:
724             seating_plan[k][j] = [-1,0]
725
726 def two_random():
727     global seating_plan
728     global highest_priority_assigned
729
730     seating_plan = [[ [random.randrange(1,10),0,assign_luggage()] for _ in range(
TWO_ROWS)] for _ in range(TWO_SEATS)]
731
732     highest_priority_assigned = 10
733     two_first_class()
734     two_cleanup()
735     naughty_people_two()
736     group_shit_two()
737
738 def two_back_to_front():
739     global seating_plan
740     global highest_priority_assigned
```

```

741 highest_priority_assigned = 0
742 #Generate an empty plane
743 seating_plan = [[ [0,0,0] for _ in range(TWO_ROWS)] for _ in range(TWO_SEATS)
744 ]
745 for _ in range(TWO_SEATS):
746     for k in range(4,TWO_A_ROWS):
747         seating_plan[_][k] = [k,0,assign_luggage()]
748
749 for _ in range(TWO_SEATS):
750     for k in (range(TWO_B_ROWS)):
751
752         seating_plan[_][(TWO_B_ROWS-k)+(TWO_A_ROWS+GAP_SIZE-1)] = [k,0,
assign_luggage()]
753
754 highest_priority_assigned = TWO_B_ROWS
755
756 two_first_class()
757 two_cleanup()
758 naughty_people_two()
759 group_shit_two()
760
761 def two_front_to_back():
762     global seating_plan
763     global highest_priority_assigned
764     highest_priority_assigned = 0
765     #Generate an empty plane
766     seating_plan = [[ [0,0,0] for _ in range(TWO_ROWS)] for _ in range(TWO_SEATS)
767 ]
768     for _ in range(TWO_SEATS):
769         for k in range(4,TWO_A_ROWS):
770             seating_plan[_][TWO_A_ROWS-k] = [k,0,assign_luggage()]
771
772     for _ in range(TWO_SEATS):
773         for k in (range(TWO_B_ROWS)):
774
775             seating_plan[_][(k)+(TWO_A_ROWS+GAP_SIZE)] = [k,0,assign_luggage()]
776
777     highest_priority_assigned = TWO_B_ROWS
778
779     two_first_class()
780     two_cleanup()
781     naughty_people_two()
782     group_shit_two()
783
784 def two_reverse_wilma_widebody():
785     global seating_plan
786     global highest_priority_assigned
787
788     highest_priority_assigned = 2
789
790     #Make an empty
791     seating_plan = [[ [-1,0,0] for _ in range(TWO_ROWS)] for _ in range(TWO_SEATS)
792 )]
793
794     for i in range(TWO_SEATS):
795         if i in [1,3,5,7]:
796             seating_plan[i]= [[2,0,assign_luggage()] for _ in range(TWO_ROWS)]
797         elif i in [0,4,8]:

```

```
797         seating_plan[i]= [[1,0,assign_luggage()] for _ in range(TWO_ROWS)]
798
799     two_first_class()
800     two_cleanup()
801     naughty_people_two()
802     group_shit_two()
803
804
805 def two_reverse_sections_360():
806     global seating_plan
807
808     global highest_priority_assigned
809
810     highest_priority_assigned = 3
811     fjuk = []
812     for i in range(TWO_SEATS):
813         aisles = []
814         for k in range(0,8):
815             aisles.append([3,0,assign_luggage()])
816         for k in range(8,13):
817             aisles.append([2,0,assign_luggage()])
818         for k in range(13,21):
819             aisles.append([1,0,assign_luggage()])
820         for k in range(21,28):
821             aisles.append([1,0,assign_luggage()])
822         for k in range(28,36):
823             aisles.append([2,0,assign_luggage()])
824         for k in range(36,TWO_ROWS):
825             aisles.append([3,0,assign_luggage()])
826         fjuk.append(aisles)
827     seating_plan = fjuk
828
829     two_first_class()
830     two_cleanup()
831     naughty_people_two()
832     group_shit_two()
833
834
835
836
837
838
839
840 #Logic
841 def check_locker_space_wide(luggage_number, current_row, seat, lockers):
842     # if passenger has no baggage
843     if luggage_number == 0:
844         return 0
845
846     sublocker = math.floor((seat)/7)
847
848     if seat % 7 < 3:
849         nbins = lockers[sublocker][current_row-1][0]
850         lockers[sublocker][current_row-1][0] -= luggage_number
851
852     elif seat % 7 > 3:
853         nbins = lockers[sublocker][current_row-1][1]
854         lockers[sublocker][current_row-1][1] -= luggage_number
855
856     # derivations in writeup
```

```

857     if luggage_number == 1:
858         t = (4)/(1-(0.8*((nbins-2)))/6)
859     if luggage_number == 2:
860         t = (4)/(1-(0.8*((nbins-2)))/6) + (2.25)/(1-((nbins-2))/6)
861
862     return t
863 def check_locker_space(luggage_number, current_row, down, lockers):
864     # if passenger has no baggage
865     if luggage_number == 0:
866         return 0
867
868     # if on right side of aisle
869     if down==True:
870
871         nbins = lockers[NUM_ROWS-current_row-1][1]
872         lockers[NUM_ROWS-current_row-1][1] -= luggage_number
873     else:
874
875         nbins = lockers[NUM_ROWS-current_row-1][0]
876         lockers[NUM_ROWS-current_row-1][0] -= luggage_number
877     if luggage_number == 1:
878         t = (4)/(1-(0.8*((nbins-2)))/6)
879     if luggage_number == 2:
880         t = (4)/(1-(0.8*((nbins-2)))/6) + (2.25)/(1-((nbins-2))/6)
881
882     return t
883 def check_locker_space_Two(luggage_number, current_row, seat, lockers,sectionA):
884
885     if sectionA:
886         thingy = 0
887     else:
888         thingy = 1
889
890     # if passenger has no baggage
891     if luggage_number == 0:
892         return 0
893
894     # if on right side of aisle
895     if seat > 3:
896
897         nbins = lockers[thingy][current_row][1]
898         lockers[thingy][current_row][1] -= luggage_number
899     else:
900
901         nbins = lockers[thingy][current_row][0]
902         lockers[thingy][current_row][0] -= luggage_number
903     if luggage_number == 1:
904         t = (4)/(1-(0.8*((nbins-2)))/6)
905     if luggage_number == 2:
906         t = (4)/(1-(0.8*((nbins-2)))/6) + (2.25)/(1-((nbins-2))/6)
907
908     return t
909 def move_up(seat, row):
910
911     if seating_plan[seat-1][row][0] == -1:
912         seating_plan[seat-1][row] = seating_plan[seat][row]
913         seating_plan[seat-1][row][1] = 0
914         seating_plan[seat][row] = [-1,0]
915 def move_down(seat, row):
916

```

```
917
918     if seating_plan[seat+1][row][0] == -1:
919         seating_plan[seat+1][row] = seating_plan[seat][row]
920         seating_plan[seat+1][row][1] = 0
921         seating_plan[seat][row] = [-1,0]
922 def aisle_take_above(row,hello,world):
923     #Move person from above into aisle
924     seating_plan[AISLE_INDEX][row] = seating_plan[AISLE_INDEX-1][row]
925     seating_plan[AISLE_INDEX-1][row] = [-1,0]
926
927     #Luggage
928     seating_plan[AISLE_INDEX][row][INTERNAL_COCK] = -locker_shit_type[
boarding_type](seating_plan[AISLE_INDEX][row][HAS_LUGGAGE], row, True, lockers
)
929 def aisle_take_below(row,frick,me):
930     #Move person from below into aisle
931     seating_plan[AISLE_INDEX][row] = seating_plan[AISLE_INDEX+1][row]
932     seating_plan[AISLE_INDEX+1][row] = [-1,0]
933
934
935     seating_plan[AISLE_INDEX][row][INTERNAL_COCK] = -locker_shit_type[
boarding_type](seating_plan[AISLE_INDEX][row][HAS_LUGGAGE], row, False,
lockers)
936 def aisle_take_above_wide(row,current_aisles,uguil):
937     #Move person from above into aisle
938     #Luggage
939     seating_plan[current_aisles-1][row][INTERNAL_COCK] = -locker_shit_type[
boarding_type](seating_plan[current_aisles-1][row][HAS_LUGGAGE], row,
current_aisles-1, lockers)
940
941     seating_plan[current_aisles][row] = seating_plan[current_aisles-1][row]
942     seating_plan[current_aisles-1][row] = [-1,0]
943 def aisle_take_above_Two(row,current_aisles,SectionA):
944     #Move person from above into aisle
945     #Luggage
946     if row != 3:
947         seating_plan[current_aisles-1][row][INTERNAL_COCK] = -locker_shit_type[
boarding_type](seating_plan[current_aisles-1][row][HAS_LUGGAGE], row-(GAP_SIZE
+TWO_A_ROWS), current_aisles-1, lockers,SectionA)
948
949     seating_plan[current_aisles][row] = seating_plan[current_aisles-1][row]
950     seating_plan[current_aisles-1][row] = [-1,0]
951
952 def aisle_take_below_Two(row,current_aisles,sectionA):
953     #Move person from below into aisle
954     if row != 3:
955         seating_plan[current_aisles+1][row][INTERNAL_COCK] = -locker_shit_type[
boarding_type](seating_plan[current_aisles+1][row][HAS_LUGGAGE], row-(GAP_SIZE
+TWO_A_ROWS), current_aisles+1, lockers,sectionA)
956
957     seating_plan[current_aisles][row] = seating_plan[current_aisles+1][row]
958     seating_plan[current_aisles+1][row] = [-1,0]
959
960 def aisle_take_below_wide(row,current_aisles,aszgasdhasdh):
961     #Move person from below into aisle
962     if row != 0:
963         seating_plan[current_aisles+1][row][INTERNAL_COCK] = -locker_shit_type[
boarding_type](seating_plan[current_aisles+1][row][HAS_LUGGAGE], row,
current_aisles+1, lockers)
964
```



```

965     seating_plan[current_aisles][row] = seating_plan[current_aisles+1][row]
966     seating_plan[current_aisles+1][row] = [-1,0]
967
968
969
970
971
972 def aisle_take_left(row, current_aisles, whythefucknot):
973
974     #Move person from right into aisle
975     seating_plan[current_aisles][row] = seating_plan[current_aisles][row-1]
976     seating_plan[current_aisles][row-1] = [-1,0]
977     seating_plan[current_aisles][row][INTERNAL_COCK] = 0
978 def aisle_take_right(row, idonot, careanymore):
979
980     #Move person from right into aisle
981     seating_plan[AISLE_INDEX][row] = seating_plan[AISLE_INDEX][row+1]
982     seating_plan[AISLE_INDEX][row+1] = [-1,0]
983     seating_plan[AISLE_INDEX][row][INTERNAL_COCK] = 0
984
985 def aisle_take_right_wide(row, current_aisles, whythefucknot):
986
987     #Move person from right into aisle
988     seating_plan[current_aisles][row] = seating_plan[current_aisles][row+1]
989     seating_plan[current_aisles][row+1] = [-1,0]
990     seating_plan[current_aisles][row][INTERNAL_COCK] = 0
991
992 #While loop
993 def off_the_plane(generation_method, text):
994     global im, fig
995     #isual shot
996
997     global seating_plan
998     global priority_weightings
999     global lockers
1000     test_cases = []
1001
1002
1003
1004
1005     for i in range(N_TEST_CASES):
1006         generation_method()
1007         total_time=0
1008         left_plane = 0
1009         priority_weightings = generate_prioities(highest_priority_assigned)
1010         lockers = bag_shit_type[boarding_type]()
1011         if VISUALIZER:
1012             im, fig = render_type[boarding_type]()
1013
1014         while True:
1015             if boarding_type == TWO:
1016                 #Exit square
1017                 if seating_plan[2][41][PRIORITY] != -1 and seating_plan[2][41][
INTERNAL_COCK] >= TIME_TO_MOVE:
1018                     #Empty square
1019                     seating_plan[2][41] = [-1,0]
1020                     left_plane += 1
1021
1022                 if seating_plan[6][41][PRIORITY] != -1 and seating_plan[6][41][
INTERNAL_COCK] >= TIME_TO_MOVE:

```

```

1023         #Empty square
1024         seating_plan[6][41] = [-1,0]
1025         left_plane += 1
1026         if seating_plan[TWO_SEATS-1][3][PRIORITY] != -1 and seating_plan[
TWO_SEATS-1][3][INTERNAL_COCK] >= TIME_TO_MOVE:
1027             #Empty square
1028             seating_plan[TWO_SEATS-1][3] = [-1,0]
1029             left_plane += 1
1030
1031
1032         #End code shot
1033         if left_plane == 249:
1034             test_cases.append(total_time)
1035             break
1036
1037         for current_aisle in (range(TWO_SEATS)):
1038             if seating_plan[current_aisle][3][PRIORITY] == -1:
1039
1040                 priorities = [0,0,0,0]
1041                 possible_moves = [aisle_take_above_Two,
aisle_take_below_Two, aisle_take_right_wide, aisle_take_left ]
1042                 total_move_possibilites = 0
1043
1044                 #Get things to check
1045                 is_person_above_moving = seating_plan[current_aisle
-1][3][PRIORITY] != -1 and seating_plan[current_aisle-1][3][INTERNAL_COCK] >=
TIME_TO_MOVE_PAST_SEAT
1046
1047                 if is_person_above_moving:
1048                     priorities[0] = priority_weightings[seating_plan[
current_aisle-1][3][PRIORITY]-1]
1049                     total_move_possibilites +=1
1050                     if current_aisle != 8:
1051                         is_person_below_moving = False #seating_plan[
current_aisle+1][3][PRIORITY] != -1 and seating_plan[current_aisle+1][3][
INTERNAL_COCK] >= TIME_TO_MOVE_PAST_SEAT
1052                     else: is_person_below_moving = 0
1053                     if is_person_below_moving:
1054
1055                         priorities[1] = 0 #priority_weightings[
seating_plan[current_aisle+1][3][PRIORITY]-1]
1056                         total_move_possibilites +=1
1057                         #Prevent indexing error
1058                         if True: #row+1+ec != TWO_ROWS:
1059                             is_person_right_moving = seating_plan[
current_aisle][3+1][PRIORITY] != -1 and seating_plan[current_aisle][3+1][
INTERNAL_COCK] >= TIME_TO_MOVE
1060                         else:
1061                             is_person_right_moving = 0
1062
1063                         if is_person_right_moving and (current_aisle == 2 or
current_aisle == 6):
1064
1065                             priorities[2] = priority_weightings[seating_plan[
current_aisle][3+1][PRIORITY]-1]
1066                             total_move_possibilites +=1
1067                             #ewginsdaogvnadsklbvasj nwklsfnwdsf
1068                             if True: #row+1+ec != TWO_ROWS:
1069                                 is_person_left_moving = seating_plan[
current_aisle][3-1][PRIORITY] != -1 and seating_plan[current_aisle][3-1][

```

```

INTERNAL_COCK] >= TIME_TO_MOVE
1070         else:
1071             is_person_right_moving = 0
1072
1073         if is_person_left_moving and (current_aisle == 2 or
current_aisle == 6) :
1074             priorities[3] = priority_weightings[seating_plan[
current_aisle][3-1][PRIORITY]-1]
1075             total_move_possibilites +=1
1076
1077             #Decide who moves above and below
1078             if total_move_possibilites > 0:
1079                 #Reset time
1080                 seating_plan[current_aisle][3][INTERNAL_COCK] = 0
1081
1082                 #frick knows what is happening here but it works
so it stays
1083                 move = numpy.argwhere(priorities == numpy.amax(
priorities))
1084                 possible_moves[(random.choice(move))[0]](3,
current_aisle, False)
1085
1086         for current_aisle in range(2, TWO_SEATS, 4):
1087
1088             for row in reversed(range(0, TWO_B_ROWS)):
1089
1090                 #extra constant
1091                 ec = GAP_SIZE+TWO_A_ROWS
1092                 #Check if aisles place is empty
1093                 if seating_plan[current_aisle][row+ec][PRIORITY] == -1:
1094
1095                     priorities = [0,0,0]
1096                     possible_moves = [aisle_take_above_Two,
aisle_take_below_Two, aisle_take_left ]
1097                     total_move_possibilites = 0
1098
1099                     #Get things to check
1100                     is_person_above_moving = seating_plan[current_aisle
-1][row+ec][PRIORITY] != -1 and seating_plan[current_aisle-1][row+ec][
INTERNAL_COCK] >= TIME_TO_MOVE_PAST_SEAT
1101
1102                     if is_person_above_moving:
1103                         priorities[0] = priority_weightings[seating_plan[
current_aisle-1][row+ec][PRIORITY]-1]
1104                         total_move_possibilites +=1
1105
1106                     is_person_below_moving = seating_plan[current_aisle
+1][row+ec][PRIORITY] != -1 and seating_plan[current_aisle+1][row+ec][
INTERNAL_COCK] >= TIME_TO_MOVE_PAST_SEAT
1107                     if is_person_below_moving:
1108
1109                         priorities[1] = priority_weightings[seating_plan[
current_aisle+1][row+ec][PRIORITY]-1]
1110                         total_move_possibilites +=1
1111                         #Prevent indexing error
1112                         if True: #row+1+ec != TWO_ROWS:
1113                             is_person_right_moving = seating_plan[
current_aisle][row-1+ec][PRIORITY] != -1 and seating_plan[current_aisle][row
-1+ec][INTERNAL_COCK] >= TIME_TO_MOVE
1114                             else:

```



```

1161         if is_person_right_moving:
1162
1163             priorities[2] = priority_weightings[
seating_plan[current_aisle][row+1][PRIORITY]-1]
1164                 total_move_possibilites +=1
1165             #ewginsdaogvnadsklbvasj nwklsfnwdsf
1166             if True: #row+1+ec != TWO_ROWS:
1167                 is_person_left_moving = seating_plan[
current_aisle][row-1][PRIORITY] != -1 and seating_plan[current_aisle][row-1][
INTERNAL_COCK] >= TIME_TO_MOVE
1168             else:
1169                 is_person_right_moving = 0
1170
1171             if is_person_left_moving and row == 3:
1172                 priorities[3] = priority_weightings[
seating_plan[current_aisle][row-1][PRIORITY]-1]
1173                 total_move_possibilites +=1
1174
1175             #Decide who moves above and below
1176             if total_move_possibilites > 0:
1177                 #Reset time
1178                 seating_plan[current_aisle][row][
INTERNAL_COCK] = 0
1179
1180                 #frick knows what is happening here but it
works so it stays
1181                 move = numpy.argwhere(priorities == numpy.
amax(priorities))
1182                 possible_moves[(random.choice(move))[0]](row,
current_aisle, False)
1183
1184             #Move towards aisle
1185             for row in range(TWO_ROWS):
1186                 if row!= 3:
1187                     if seating_plan[0][row][INTERNAL_COCK] >=
TIME_TO_MOVE_PAST_SEAT:
1188                         move_down(0, row)
1189                     if seating_plan[TWO_SEATS-1][row][INTERNAL_COCK] >=
TIME_TO_MOVE_PAST_SEAT and seating_plan[TWO_SEATS-1][row][PRIORITY] != -1:
1190                         move_up(TWO_SEATS-1, row)
1191                     if seating_plan[4][row][INTERNAL_COCK] >=
TIME_TO_MOVE_PAST_SEAT:
1192                         if random.randint(0,1) and seating_plan[3][row][
PRIORITY] == -1:
1193                             move_up(4, row)
1194                         elif seating_plan[5][row][PRIORITY] == -1:
1195                             move_down(4, row)
1196
1197
1198
1199
1200
1201                 for i in range(TWO_SEATS):
1202                     for k in range(TWO_ROWS):
1203                         seating_plan[i][k][INTERNAL_COCK] += TIME_STEP
1204
1205
1206
1207
1208

```

```

1209
1210
1211
1212     elif boarding_type == WIDEBODY:
1213         #Exit square
1214         if seating_plan[0][0][PRIORITY] != -1 and seating_plan[0][0][
INTERNAL_COCK] >= TIME_TO_MOVE:
1215             #Empty square
1216             seating_plan[0][0] = [-1,0]
1217
1218             #End code shot
1219             left_plane += 1
1220             if left_plane == (WIDE_WING_ROWS-1)*(WIDE_WING_SEATS-4)-18:
1221                 test_cases.append(total_time)
1222
1223                 break
1224
1225         #Down queue do not touch 2am code
1226         for seat in range(WIDE_WING_SEATS):
1227
1228             #if can move something in
1229             if seating_plan[seat][0][PRIORITY] == -1:
1230
1231                 priorities = [0,0]
1232                 possible_moves = [ aisle_take_below_wide ,
aisle_take_right_wide ]
1233                 total_move_possibilites = 0
1234
1235                 #Check shot
1236                 is_person_below_moving = seating_plan[seat][1][PRIORITY]
!= -1 and seating_plan[seat][1][INTERNAL_COCK] >= TIME_TO_MOVE_PAST_SEAT and
seat in [3,10,17,24]
1237                 if is_person_below_moving:
1238
1239                     priorities[1] = priority_weightings[seating_plan[seat
] [1][PRIORITY]-1]
1240                     total_move_possibilites +=1
1241                     #Prevent indexing error
1242                     if seat != WIDE_WING_SEATS-1:
1243                         is_person_right_moving = seating_plan[seat+1][0][
PRIORITY] != -1 and seating_plan[seat+1][0][INTERNAL_COCK] >= TIME_TO_MOVE
1244                     else:
1245                         is_person_right_moving = 0
1246
1247                     if is_person_right_moving:
1248
1249                         priorities[0] = priority_weightings[seating_plan[seat
+1][0][PRIORITY]-1]
1250                         total_move_possibilites +=1
1251
1252                     #Decide who moves above and below
1253                     if total_move_possibilites > 0:
1254                         #Reset time
1255                         seating_plan[seat][0][INTERNAL_COCK] = 0
1256
1257                         #frick knows what is happening here but it works so
it stays
1258                         move = numpy.argwhere(priorities == numpy.amax(
priorities))
1259                         possible_moves [(random.choice(move))[0]](0, seat, False

```

```

)
1260
1261
1262
1263     for current_aisle in range(3, WIDE_WING_SEATS,7):
1264
1265
1266
1267         for row in range(1,WIDE_WING_ROWS):
1268
1269
1270
1271             #Check if aisles place is empty
1272             if seating_plan[current_aisle][row][PRIORITY] == -1:
1273
1274                 priorities = [0,0,0]
1275                 possible_moves = [aisle_take_above_wide,
aisle_take_below_wide, aisle_take_right_wide ]
1276                 total_move_possibilites = 0
1277
1278                 #Get things to check
1279                 is_person_above_moving = seating_plan[current_aisle
-1][row][PRIORITY] != -1 and seating_plan[current_aisle-1][row][INTERNAL_COCK]
>= TIME_TO_MOVE_PAST_SEAT
1280
1281                 if is_person_above_moving:
1282                     priorities[0] = priority_weightings[seating_plan[
current_aisle-1][row][PRIORITY]-1]
1283                     total_move_possibilites +=1
1284
1285                     is_person_below_moving = seating_plan[current_aisle
+1][row][PRIORITY] != -1 and seating_plan[current_aisle+1][row][INTERNAL_COCK]
>= TIME_TO_MOVE_PAST_SEAT
1286                     if is_person_below_moving:
1287
1288                         priorities[1] = priority_weightings[seating_plan[
current_aisle+1][row][PRIORITY]-1]
1289                         total_move_possibilites +=1
1290                         #Prevent indexing error
1291                         if row != WIDE_WING_ROWS-1:
1292                             is_person_right_moving = seating_plan[
current_aisle][row+1][PRIORITY] != -1 and seating_plan[current_aisle][row+1][
INTERNAL_COCK] >= TIME_TO_MOVE
1293                         else:
1294                             is_person_right_moving = 0
1295
1296                         if is_person_right_moving:
1297
1298                             priorities[2] = priority_weightings[seating_plan[
current_aisle][row+1][PRIORITY]-1]
1299                             total_move_possibilites +=1
1300
1301                         #Decide who moves above and below
1302                         if total_move_possibilites > 0:
1303                             #Reset time
1304                             seating_plan[current_aisle][row][INTERNAL_COCK] =
0
1305
1306                             #frick knows what is happening here but it works
so it stays

```

```

1307         move = numpy.argwhere(priorities == numpy.amax(
priorities))
1308         possible_moves[(random.choice(move))[0]](row,
current_aisle, False)
1309
1310
1311         # I fricking HATE INDENDATION
1312
1313
1314         #Get total amount of move possibilites
1315         #total_move_possibilites = is_person_above_moving +
is_person_below_moving + is_person_right_moving
1316
1317
1318         for current_aisle in range(0, WIDE_WING_SEATS, 7):
1319
1320
1321         # Move down
1322         for seat in reversed(range(0, 2)):
1323             # Loops through all 37 rows
1324
1325             #Move towards aisle
1326             for row in range(0, WIDE_WING_ROWS):
1327                 if seating_plan[seat+current_aisle][row][
INTERNAL_COCK] >= TIME_TO_MOVE_PAST_SEAT:
1328                     move_down(seat+current_aisle, row)
1329
1330                     #Incrasese internal clock
1331
1332
1333         # Move up
1334         for seat in range(5, 7):
1335             # Loops through all 37 rows
1336
1337             #Move towards aisle
1338             for row in range(0, WIDE_WING_ROWS):
1339                 if seating_plan[seat+current_aisle][row][
INTERNAL_COCK] >= TIME_TO_MOVE_PAST_SEAT:
1340                     move_up(seat+current_aisle, row)
1341                     #Incrasese internal clock
1342         for i in range(WIDE_WING_SEATS):
1343             for k in range(WIDE_WING_ROWS):
1344                 seating_plan[i][k][INTERNAL_COCK] += TIME_STEP
1345
1346
1347         elif boarding_type == NORMAL:
1348
1349             #Exit square
1350             if seating_plan[AISLE_INDEX][0][PRIORITY] != -1 and seating_plan
[3][0][INTERNAL_COCK] >= TIME_TO_MOVE:
1351                 #Empty square
1352                 seating_plan[AISLE_INDEX][0] = [-1, 0]
1353
1354             #End code shot
1355             left_plane += 1
1356             if left_plane == NUM_ROWS*NUM_SEATS:
1357                 test_cases.append(total_time)
1358
1359             break
1360

```



```

1361
1362
1363     #Aisle handling code
1364     for row in range(0, NUM_ROWS):
1365         #Check if aisles place is empty
1366         if seating_plan[AISLE_INDEX][row][PRIORITY] == -1:
1367
1368
1369             priorities = [0,0,0]
1370             possible_moves = [aisle_take_above, aisle_take_below,
aisle_take_right ]
1371             total_move_possibilites = 0
1372
1373             #Get things to check
1374             is_person_above_moving = seating_plan[AISLE_INDEX-1][row
][PRIORITY] != -1 and seating_plan[AISLE_INDEX-1][row][INTERNAL_COCK] >=
TIME_TO_MOVE_PAST_SEAT
1375
1376             if is_person_above_moving:
1377                 priorities[0] = priority_weightings[seating_plan[
AISLE_INDEX-1][row][PRIORITY]-1]
1378                 total_move_possibilites +=1
1379
1380             is_person_below_moving = seating_plan[AISLE_INDEX+1][row
][PRIORITY] != -1 and seating_plan[AISLE_INDEX+1][row][INTERNAL_COCK] >=
TIME_TO_MOVE_PAST_SEAT
1381             if is_person_below_moving:
1382
1383                 priorities[1] = priority_weightings[seating_plan[
AISLE_INDEX+1][row][PRIORITY]-1]
1384                 total_move_possibilites +=1
1385
1386             #Prevent indexing error
1387             if row != NUM_ROWS-1:
1388                 is_person_right_moving = seating_plan[AISLE_INDEX][
row+1][PRIORITY] != -1 and seating_plan[AISLE_INDEX][row+1][INTERNAL_COCK] >=
TIME_TO_MOVE
1389             else:
1390                 is_person_right_moving = 0
1391
1392             if is_person_right_moving:
1393
1394                 priorities[2] = priority_weightings[seating_plan[
AISLE_INDEX][row+1][PRIORITY]-1]
1395                 total_move_possibilites +=1
1396
1397
1398             #Get total amount of move possibilites
1399             #total_move_possibilites = is_person_above_moving +
is_person_below_moving + is_person_right_moving
1400
1401             #Decide who moves above and below
1402             if total_move_possibilites > 0:
1403                 #Reset time
1404                 seating_plan[AISLE_INDEX][row][INTERNAL_COCK] = 0
1405
1406                 #frick             knows what is happening here but it
works so it stays
1407                 move = numpy.argwhere(priorities == numpy.amax(
priorities))

```

```

1408             possible_moves[(random.choice(move))[0]](row,False,
False)
1409
1410
1411
1412         # Move down
1413         for seat in reversed(range(0,2)):
1414             # Loops through all 37 rows
1415
1416             #Move towards aisle
1417             for row in range(0,NUM_ROWS):
1418                 if seating_plan[seat][row][INTERNAL_COCK] >=
TIME_TO_MOVE_PAST_SEAT:
1419                     move_down(seat,row)
1420                     #Incrasese internal clock
1421
1422
1423         # Move up
1424         for seat in range(5,7):
1425             # Loops through all 37 rows
1426
1427             #Move towards aisle
1428             for row in range(0,NUM_ROWS):
1429                 if seating_plan[seat][row][INTERNAL_COCK] >=
TIME_TO_MOVE_PAST_SEAT:
1430                     move_up(seat,row)
1431                     #Incrasese internal clock
1432
1433         for i in range(NUM_SEATS+1):
1434             for k in range(NUM_ROWS):
1435                 seating_plan[i][k][INTERNAL_COCK] += TIME_STEP
1436
1437
1438
1439
1440
1441         total_time += TIME_STEP
1442
1443         #Update render comment out if not using
1444         if VISUALIZER: update_render(seating_plan)
1445         print(text + str(sum(test_cases)/len(test_cases)))
1446         rows.append(test_cases)
1447         return test_cases
1448 #Types
1449 render_type = [intalize_render,intalize_render_widebody,intalize_render_two_thing
]
1450 bag_shit_type = [bag_shit, bag_shit_wide,bag_shit_Two]
1451 locker_shit_type = [check_locker_space,check_locker_space_wide,
check_locker_space_Two]
1452 #What thing to do
1453 NORMAL = 0
1454 WIDEBODY = 1
1455 TWO = 2
1456 #Test stuff
1457 N_TEST_CASES = 50
1458 VISUALIZER = True
1459 TIME_STEP = 0
1460 boarding_type = TWO
1461 #Data csv
1462 import csv

```

```
1463 fields = []
1464 rows = []
1465 index = []
1466 #Add the indexing
1467 for i in range(N_TEST_CASES):
1468     index.append(i)
1469
1470 rows.append(index)
1471 #Vroom
1472 seating_plan = []
1473
1474 '''
1475 #Flat body
1476 off_the_plane(random_deboard, 'Random: ')
1477 off_the_plane(sections, 'Sections: ')
1478 off_the_plane(reverse_wilma, 'Reverse Wilma: ')
1479 off_the_plane(generate_front_to_back, 'Front to back Row: ')
1480 off_the_plane(back_to_front, 'Back to Front Row: ')
1481 # field names add whatever field names that you are creating data for
1482 fields = ['Index','Random Group Adjusted', 'Front to back - sections', 'Reverse
1483     Wilma', 'Front to Back Row', 'Back to Front Row']
1484
1485
1486 #Wide body
1487 off_the_plane(random_deboard_widebody, 'Random: ')
1488 off_the_plane(sections_widebody, 'Sections: ')
1489 off_the_plane(reverse_wilma_widebody, 'Reverse Wilma: ')
1490 off_the_plane(generate_front_to_back_widebody, 'Front seat to back seat: ')
1491 off_the_plane(back_to_front_widebody, 'Back seat to Front seat: ')
1492 off_the_plane(across_widebody, 'Across: ')
1493 # field names add whatever field names that you are creating data for
1494 fields = ['Index','Random Group Adjusted', 'Front to back - sections', 'Reverse
1495     Wilma', 'Front to Back Row', 'Back to Front Row', 'Across']
1496
1497 # field names add whatever field names that you are creating data for
1498 fields = ['Index','Back to front', 'Sections', 'Random', 'Reverse Wilma', 'Front
1499     to back']
1500 off_the_plane(two_reverse_sections_360, 'back to front')
1501 #off_the_plane(two_reverse_sections_360, 'Sections')
1502 #
1503 #off_the_plane(two_front_to_back, 'front to back')
1504 #off_the_plane(two_reverse_wilma_widebody, 'reverse wilma')
1505 #off_the_plane(two_random, 'random')
1506 file_name='twoaisles.csv'
1507
1508 '''
1509 nbsensitivity = []
1510
1511 for i in range(0,41):
1512     NAUGHTY_BOY_COEFFICIENT = (i*2.5)/100
1513
1514     # put method wanted in here
1515     nbsensitivity.append(off_the_plane(back_to_front, 'back to front: '))
1516
1517     print('for test with NB coefficient {}'.format((i*2.5)/100))
1518 '''
1519 #Makes the shotinto colums honestly magic
```

```

1520 rows = zip(*rows)
1521 #Create the rows
1522 with open(file_name, 'w', newline='') as f:
1523
1524     # using csv.writer method from CSV package
1525     write = csv.writer(f)
1526
1527     write.writerow(fields)
1528
1529     write.writerows(rows)
1530 '''

```

Appendix F

```

1 import random
2 import matplotlib.pyplot as plt
3 import numpy
4 import math
5
6 # visualizer things
7
8 # render stuff that I don't understand
9 def intalize_render():
10
11     global plane
12
13     #Absolute mess of code
14     image = []
15     for i in range(NUM_SEATS+len(AISLES)):
16         subimage = []
17         for k in range(NUM_ROWS):
18             if k % 2 == 0:
19                 subimage.append(-1)
20             else:
21                 subimage.append(0)
22
23         image.append(subimage)
24
25
26     fig,ax = plt.subplots(1,1)
27     plt.set_cmap('OrRd')
28     print(image)
29     image = numpy.array(image)
30
31     im = ax.imshow(image)
32     number_of_runs = range(1,NUM_ROWS) # use your actual number_of_runs
33     ax.set_xticks(number_of_runs, minor=False)
34     ax.xaxis.grid(True, which='major')
35
36
37
38
39     ax.set_yticks(numpy.arange(0.5, NUM_SEATS+len(AISLES)+.5, 1).tolist(), minor=
False)
40     ax.yaxis.grid(True, which='major')
41
42     if plane == 'wide wing':
43         ax.set_yticklabels(['A','B','C','Aisle','D','E','F','G','H','I','Aisle','
J','K','L','M','N','O','Aisle','P','Q','R','S','T','U','Aisle','V','W','X'])

```

```
44     elif plane == 'narrow body':
45         ax.set_yticklabels(['Row A', 'Row B', 'Row C', 'Aisle', 'Row D', 'Row E', 'Row
46 F'])
47     elif plane == 'two entrance two aisle':
48         ax.set_yticklabels(['Row A', 'Row B', 'Aisle', 'Row C', 'Row D', 'Row E',
49 Aisle', 'Row F', 'Row G'])
50     ax.set_ylim(top=-0.5)
51
52     ax.set_xticks(numpy.arange(0.5, NUM_ROWS+0.5, 1).tolist(), minor=False)
53     ax.xaxis.grid(True, which='major')
54
55     xticklist = []
56     #Create list of numbers between
57     for i in range(NUM_ROWS):
58         if ((i+1) % 5 == 0) and (i != 0):
59             xticklist.append(str(i+1))
60         else:
61             xticklist.append('')
62
63     ax.set_xticklabels(xticklist)
64     ax.set_xlim(left=-0.5)
65
66     return im,fig
67 def update_render(seat_plan):
68
69     visualizer = []
70     for i,column in enumerate(seat_plan):
71         visualizer.append([])
72         for seat in column:
73             if i not in AISLES:
74                 if seat != -1:
75                     visualizer[i].append(0)
76                 else: visualizer[i].append(-1)
77             else:
78                 if seat != '':
79                     visualizer[i].append(0)
80                 else: visualizer[i].append(-1)
81
82
83     im.set_data(visualizer)
84     fig.canvas.draw_idle()
85     plt.pause(0.01)
86
87
88
89
90
91
92
93
94
95 # -----
96 # stuff to board the plane with (given a boarding queue)
97 # -----
98
99 # calculate time taken to get to seat if someone in the way
100 def get_past_people(seating_plan, passenger, current_row):
101
```

```

102 # number of people blocking seats
103 N=0
104 time_to_stop_blocking_aisle = 0
105
106
107 # aisle seat
108 for aisle in AISLES:
109     if abs(passenger[1] - aisle) == 1:
110         time_to_stop_blocking_aisle += TIME_TO_MOVE_PAST_SEAT
111 # middle or window seat: people are in the way
112 else:
113
114     for aisle in AISLES:
115
116         if passenger[1]-aisle == -3:
117
118
119             # if aisle seat taken IMPORTANT to check aisle seat first so f is
120             maximised
121             if seating_plan[passenger[1]+1][NUM_ROWS-current_row-1] != -1:
122                 N+=1
123                 f=1
124             # if middle seat taken
125             if seating_plan[passenger[1]+2][NUM_ROWS-current_row-1] != -1:
126                 N+=1
127                 f=2
128
129             break
130
131         elif passenger[1]-aisle == -2:
132
133             # if aisle seat taken
134             if seating_plan[passenger[1]+1][NUM_ROWS-current_row-1] != -1:
135                 N+=1
136                 f=1
137
138             # window seat F
139             elif passenger[1]-aisle == 3:
140                 # if aisle seat taken IMPORTANT to check aisle seat first so f is
141                 maximised
142                 if seating_plan[passenger[1]-2][NUM_ROWS-current_row-1] != -1:
143                     N+=1
144                     f=1
145                 # if middle seat taken
146                 if seating_plan[passenger[1]-1][NUM_ROWS-current_row-1] != -1:
147                     N+=1
148                     f=2
149
150             # middle seat B
151             elif passenger[1] == 2:
152                 # if aisle seat taken
153                 if seating_plan[passenger[1]-1][NUM_ROWS-current_row-1] != -1:
154                     N+=1
155                     f=1
156
157             if N==0:
158                 time_to_stop_blocking_aisle = TIME_TO_MOVE_PAST_SEAT
159             else:
160                 time_to_stop_blocking_aisle += TIME_TO_SIT_OR_STAND +

```

```

TIME_TO_MOVE_PAST_SEAT*(N+f+1)
160
161     return time_to_stop_blocking_aisle, N
162 # stow in overhead lockers
163 def check_locker_space(passenger, current_row, lockers, passengers_loaded_bags,
164     aisle=0):
165     # if passenger has no baggage
166     if passenger[2] == 0:
167         return 0
168
169     # if on right side of aisle
170
171     for aisle in AISLES:
172
173         if abs(passenger[1]-aisle) <= 3:
174
175             correct_aisle = aisle
176             break
177
178
179     if passenger[1] > correct_aisle:
180         if [passenger[0],passenger[1]] not in passengers_loaded_bags:
181             nbins = lockers[AISLES.index(correct_aisle)][NUM_ROWS-current_row
182 -1][1]
183             lockers[AISLES.index(correct_aisle)][NUM_ROWS-current_row-1][1] +=
184 passenger[2]
185             passengers_loaded_bags.append([passenger[0],passenger[1]])
186         else:
187             nbins = lockers[AISLES.index(correct_aisle)][NUM_ROWS-current_row
188 -1][0]-passenger[2]
189         else:
190             if [passenger[0],passenger[1]] not in passengers_loaded_bags:
191                 nbins = lockers[AISLES.index(correct_aisle)][NUM_ROWS-current_row
192 -1][0]
193                 lockers[AISLES.index(correct_aisle)][NUM_ROWS-current_row-1][0] +=
194 passenger[2]
195                 passengers_loaded_bags.append([passenger[0],passenger[1]])
196             else:
197                 nbins = lockers[AISLES.index(correct_aisle)][NUM_ROWS-current_row
198 -1][0]-passenger[2]
199     # derivations in writeup
200
201     if passenger[2] == 1:
202         t = (4)/(1-(0.8*nbins)/6)
203     if passenger[2] == 2:
204         t = (4)/(1-(0.8*nbins)/6) + (2.25)/(1-(nbins+1)/6)
205
206     return t
207
208 # board the plane
209 def board_the_plane(boardingQueue, family=False):
210     # initialize seating plan, top queue (if multiple aisles) and overhead
211     lockers
212     seating_plan = [[-1 for _ in range(NUM_ROWS)] for _ in range(NUM_SEATS + len
213 (AISLES))]
214     for aisle in AISLES:
215         seating_plan[aisle]=['' for _ in range(NUM_ROWS)]

```

```

210 top_queue = ['' for _ in range(NUM_SEATS+len(AISLES))]
211 lockers = [[0,0] for i in range(NUM_ROWS)] for j in range(len(AISLES))]
212 seated = []
213 passengers_loaded_bags = []
214
215 n_passengers = len(boardingQueue)
216
217 total_time=0
218
219 # this is false for all scenarios except where families are prioritized
220 if family == False:
221     time_to_move = TIME_TO_MOVE
222 else:
223     time_to_move = FAMILY_TIME_TO_MOVE
224
225 while True:
226
227     #print(seating_plan)
228
229     # loop through top queue
230     for current_column, passenger in enumerate(reversed(top_queue)):
231
232         if passenger != '':
233
234             # increase internal clock
235             passenger[3] += TIME_STEP
236
237             # check if passenger in right aisle and thus they can seat
238             if (NUM_SEATS+len(AISLES)-current_column-1) in AISLES and abs(
passenger[1]-(NUM_SEATS+len(AISLES)-current_column-1))<=AISLES[0]+1:
239
240                 # move into aisle
241                 if seating_plan[NUM_SEATS+len(AISLES)-current_column-1][0] ==
'' and passenger[3] >= time_to_move:
242                     #reset internal clock
243                     passenger[3]=0
244                     seating_plan[NUM_SEATS+len(AISLES)-current_column-1][0] =
passenger
245                     top_queue[NUM_SEATS+len(AISLES)-current_column-1]=''
246
247                 else:
248                     # if passenger in front has moved
249                     if top_queue[NUM_SEATS+len(AISLES)-current_column] == '' and
passenger[3] >= time_to_move:
250                         # move people along
251                         top_queue[NUM_SEATS+len(AISLES)-current_column] =
passenger
252                         top_queue[NUM_SEATS+len(AISLES)-current_column-1] = ''
253
254                     # reset internal clock
255                     passenger[3] = 0
256
257             for aisle in AISLES:
258
259                 # loop through aisle from back to front
260                 for current_row,passenger in enumerate(reversed(seating_plan[aisle]))
:
261
262                     if passenger != '':
263

```



```

264         # increase internal clock
265         #print(seating_plan[aisle])
266         passenger[3] += TIME_STEP
267
268         # check if passenger in right row and thus they can seat
269         if passenger[0] == NUM_ROWS - current_row:
270
271             # if passenger has baggage
272             time_to_stow = check_locker_space(passenger, current_row,
lockers ,passengers_loaded_bags)
273
274
275             # time it takes to stop blocking aisle and number of
people in the way
276             try:
277                 time_to_stop_blocking_aisle=passenger[5]
278             except:
279                 time_to_stop_blocking_aisle, N = get_past_people(
seating_plan, passenger, current_row)
280                 passenger.append(time_to_stop_blocking_aisle)
281
282
283             # make sure there is an empty space
284             if N==2 and current_row != 0 and seating_plan[aisle][
NUM_ROWS-current_row] != '' and current_row != 0:
285                 time_to_wait_for_spot_in_aisle += time_to_move -
passenger[3]
286             else:
287                 time_to_wait_for_spot_in_aisle=0
288
289
290             # if time to wait has finished i.e. SIT DOWN BE HUMBLE
291             if passenger[3] >= time_to_stop_blocking_aisle +
time_to_stow + time_to_wait_for_spot_in_aisle:
292
293                 seating_plan[passenger[1]][passenger[0]-1] =
passenger
294                 seated.append(passenger)
295                 #print(seated)
296
297                 # set queue place to empty
298                 seating_plan[aisle][NUM_ROWS-current_row-1]=''
299
300
301
302             else:
303                 # if passenger in front has moved
304                 if seating_plan[aisle][NUM_ROWS-current_row] == '' and
passenger[3] >= time_to_move:
305                     # move people along
306                     seating_plan[aisle][NUM_ROWS-current_row] = passenger
307                     seating_plan[aisle][NUM_ROWS-current_row-1] = ''
308
309                     # reset internal clock
310                     seating_plan[aisle][NUM_ROWS - current_row][3] = 0
311
312             if VISUALIZER: update_render(seating_plan)
313
314
315         total_time += TIME_STEP

```

```

316         if len(seated) == n_passengers:
317             return total_time
318
319
320
321         if top_queue[0] == '' and len(boardingQueue)!=0:
322
323
324             # only considered in method where families board first.
325             if family == True and boardingQueue[0] == 'b':
326                 time_to_move = NON_FAMILY_TIME_TO_MOVE
327                 boardingQueue.pop(0)
328
329             #Set first place in isle to the first passenger in the seat data
330             seating_plan[3] then remove it from seat data
331             top_queue[0] = boardingQueue[0]
332             boardingQueue.pop(0)
333
334 # luggage
335 def assign_luggage():
336     return random.choices([0,1,2], weights=BAG_COEFFICIENT, k=1)[0]
337
338 # naughty boy
339 def is_not_disobedient():
340     return random.randrange(100) > NAUGHTY_BOY_COEFFICIENT*100
341 # create a group size
342 def group_size(group_weights):
343
344     return random.choices([1,2,3], weights=group_weights, k=1)[0]
345
346 # return average of list
347 def average(x):
348     return sum(x)/len(x)
349
350
351
352
353
354
355
356 # create order of boarding
357 def create_boarding_order_for_section_but_with_groups(boarding_section,
358 other_section1, other_section2, start_row, end_row):
359     current_group_member = 0
360     current_group_section = 1
361     current_group_size = 1
362     boarding_section.append([])
363
364     for row in range(start_row, end_row+1):
365         for seat in range(0, NUM_SEATS+len(AISLES)):
366
367             if seat not in AISLES:
368
369                 current_group_member += 1
370
371                 if current_group_section == 1:
372                     boarding_section[-1].append([row, seat, assign_luggage(), 0])
373                 elif current_group_section == 2:
374                     other_section1[-1].append([row, seat, assign_luggage(), 0])

```

```
374         elif current_group_section == 3:
375             other_section2[-1].append([row, seat, assign_luggage(), 0])
376
377
378         if current_group_member == current_group_size:
379
380             for aisle in AISLES:
381
382                 if seat-aisle in [2,3]:
383                     if current_group_section == 1:
384                         boarding_section[-1].reverse()
385                     elif current_group_section == 2:
386                         other_section1[-1].reverse()
387                     elif current_group_section == 3:
388                         other_section2[-1].reverse()
389
390                 if seat-aisle in [-3,-2,1]:
391                     current_group_size = group_size((
392 SINGLE_PRINGLE_COEFFICIENT, COUPLES_COEFFICIENT, THREESOME_COEFFICIENT))
393                 elif seat-aisle in [-1,2]:
394                     current_group_size = group_size((
395 SINGLE_PRINGLE_COEFFICIENT, COUPLES_COEFFICIENT, 0))
396                 elif seat == 6:
397                     current_group_size = 1
398
399                 current_group_member = 0
400
401
402                 if is_not_disobedient():
403                     current_group_section = 1
404                     boarding_section.append([])
405
406                 # else they try board during different sections
407                 else:
408                     if random.randrange(100) < 50:
409                         current_group_section = 2
410                         other_section1.append([])
411                     else:
412                         current_group_section = 3
413                         other_section2.append([])
414
415
416
417
418
419
420 # create order of boarding for doing windows first
421 def create_boarding_order_for_aisle(boarding_section, other_section1,
422 other_section2, seats):
423     for seat in seats:
424         for row in range(1, NUM_ROWS+1):
425
426             # if passenger is not useless
427             if is_not_disobedient():
428                 boarding_section.append([row, seat, assign_luggage(), 0])
429             # else they try board during different sections
430             else:
```

```
431         if random.randrange(100) < 50:
432             other_section1.append([row, seat, assign_luggage(), 0])
433         else:
434             other_section2.append([row, seat, assign_luggage(), 0])
435
436 # create order of boarding for doing windows first using groups
437 def create_boarding_order_for_aisle_but_with_groups(boarding_section,
438             other_section1, other_section2, seats):
439
440     for seat in seats:
441         # window seats
442         for row in range(1, NUM_ROWS+1):
443
444
445         # check if item in group already appended
446         if (not any([row, seat] in x for x in boarding_section)
447             and not any([row, seat] in x for x in other_section1)
448             and not any([row, seat] in x for x in other_section2)):
449
450             # if passenger is not useless
451             if is_not_disobedient():
452
453                 for aisle in AISLES:
454
455                     if aisle-seat == 3:
456
457                         current_group_size = group_size((70,50,20))
458
459                         if current_group_size == 3:
460                             boarding_section.append([[row, seat], [row, seat
461 +1], [row, seat+2]])
462
463                         elif current_group_size == 2:
464                             boarding_section.append([[row, seat], [row, seat
465 +1]])
466
467                         else:
468                             boarding_section.append([[row, seat]])
469
470                     elif aisle-seat == -3:
471                         current_group_size = group_size((70,50,20))
472
473                         if current_group_size == 3:
474                             boarding_section.append([[row, seat], [row, seat
475 -1], [row, seat-2]])
476
477                         elif current_group_size == 2:
478                             boarding_section.append([[row, seat], [row, seat
479 -1]])
480
481                         else:
482                             boarding_section.append([[row, seat]])
483
484                     elif aisle-seat == 2:
485                         current_group_size = group_size((80,40,0))
486                         if current_group_size == 2:
487                             boarding_section.append([[row, seat], [row, seat
488 +1]])
489
490                         else:
491                             boarding_section.append([[row, seat]])
492
493                     elif aisle-seat == -2:
494                         current_group_size = group_size((80,40,0))
495                         if current_group_size == 2:
496                             boarding_section.append([[row, seat], [row, seat
```

```

-1]])
485         else:
486             boarding_section.append([[row, seat]])
487
488         else: boarding_section.append([[row, seat]])
489
490         break
491
492     # else they try board during different sections
493     else:
494         if random.randrange(100) < 50:
495             other_section1.append([[row, seat]])
496         else:
497             other_section2.append([[row, seat]])
498
499
500
501 # reduce boarding queue capacity due to Covid
502 def cull_boarding_queue(boarding_queue):
503     #this function has two aims: reduce capacity due to COVID, and remove any
504     #seats not included in planes
505
506     # first see if need to cull the seats that would be in grid of planes, but
507     # not there
508     # remove them here as easier than having to not add them in the first place
509     # in every method
510     global plane
511     if plane == 'wide wing':
512         for index,passenger in enumerate(boarding_queue):
513             # Seats A B C V W X in rows 1-3
514             if passenger[0]-1 in [0,1,2] and passenger[1] in [0,1,2,25,26,27]:
515                 del boarding_queue[index]
516     elif plane == 'narrow body':
517         for index,passenger in enumerate(boarding_queue):
518             # Row 1 seats D E F
519             if passenger[0]-1 in [0] and passenger[1] in [4,5,6]:
520                 del boarding_queue[index]
521     if COVID_CAPACITY==0:
522         return boarding_queue
523     target_to_kill = math.floor((COVID_CAPACITY)*NUM_SEATS)
524     for row in range(NUM_ROWS):
525
526         killed = 0
527         for index,passenger in enumerate(boarding_queue):
528             if passenger[0] == row:
529                 killed += 1
530                 del boarding_queue[index]
531             if killed==target_to_kill:
532                 break
533     return boarding_queue
534
535 # -----
536 # BOARDING METHODS
537 # -----
538
539 # boarding in random order
540 def random_boarding():

```

```

541 test_cases = []
542 for _ in range(N_TEST_CASES):
543     boardingQueue = []
544     for row in range(1, NUM_ROWS+1):
545         for seat in range(NUM_SEATS+len(AISLES)):
546
547             # assign bag based on probability that passenger has bag
548             if seat not in AISLES: boardingQueue.append([row, seat,
assign_luggage(), 0])
549
550         random.shuffle(boardingQueue)
551
552         test_cases.append(board_the_plane(boardingQueue, AISLES))
553
554     print('Random: ', sum(test_cases)/len(test_cases))
555
556
557
558
559 def random_boarding_with_groups():
560
561     test_cases = []
562     for _ in range(N_TEST_CASES):
563         boardingQueue = [[]]
564
565         current_group_member=0
566         current_group_size = group_size((SINGLE_PRINGLE_COEFFICIENT,
COUPLES_COEFFICIENT, THREESOME_COEFFICIENT))
567
568         for row in range(1, NUM_ROWS+1):
569
570
571             for seat in range(0, NUM_SEATS+len(AISLES)):
572
573                 if seat not in AISLES: boardingQueue[-1].append([row, seat,
assign_luggage(), 0])
574
575                 current_group_member += 1
576
577                 if current_group_member == current_group_size:
578
579                     for aisle in AISLES:
580
581                         if seat-aisle in [2,3]:
582                             boardingQueue[-1].reverse()
583
584
585                         if seat-aisle in [-3,-2,1]:
586                             current_group_size = group_size((
SINGLE_PRINGLE_COEFFICIENT, COUPLES_COEFFICIENT, THREESOME_COEFFICIENT))
587                         elif seat-aisle in [-1,2]:
588                             current_group_size = group_size((
SINGLE_PRINGLE_COEFFICIENT, COUPLES_COEFFICIENT, 0))
589                         elif seat == 6:
590                             current_group_size = 1
591
592                             current_group_member = 0
593                             boardingQueue.append([])
594
595                             break

```

```
596
597
598
599     random.shuffle(boardingQueue)
600
601     # flatten groups
602
603     boardingQueue = [j for sub in boardingQueue for j in sub]
604
605     #print(boardingQueue)
606
607     boardingQueue = cull_boarding_queue(boardingQueue)
608
609     test_cases.append(board_the_plane(boardingQueue))
610
611     print('Random with groups: ', sum(test_cases)/len(test_cases))
612     print(test_cases)
613     return average(test_cases)
614
615
616 # sectional boarding but with groups
617 def section_boarding_with_groups():
618
619     test_cases = []
620     amf, fma = [], []
621     for _ in range(N_TEST_CASES):
622
623         aft, middle, front = [], [], []
624
625         # aft section
626         create_boarding_order_for_section_but_with_groups(aft, middle, front,
627 A_SEC_START, A_SEC_END)
628         # middle section
629         create_boarding_order_for_section_but_with_groups(middle, aft, front,
630 M_SEC_START, M_SEC_END)
631         # front section
632         create_boarding_order_for_section_but_with_groups(front, middle, aft,
633 F_SEC_START, F_SEC_END)
634
635         random.shuffle(aft)
636         random.shuffle(middle)
637         random.shuffle(front)
638
639         #print(boardingQueue)
640         boardingQueue = aft+middle+front
641         boardingQueue = [j for sub in boardingQueue for j in sub]
642         boardingQueue = cull_boarding_queue(boardingQueue)
643         amf.append(board_the_plane(boardingQueue))
644         #boardingQueue = front+middle+aft
645         #boardingQueue = [j for sub in boardingQueue for j in sub]
646         #fma.append(board_the_plane(boardingQueue))
647
648     print('Sectional amf: ', average(amf))
649     #print('Sectional fma: ', average(fma))
650
651     return(average(amf))
652
```

```
653
654
655
656 # boarding by seat but allowing groups to board together
657 def seat_boarding_with_groups():
658
659     test_cases = []
660     boardingQueue=[]
661     for _ in range(N_TEST_CASES):
662
663         window,middle,aisle = [],[],[]
664
665         # window seats
666         #window_seats = [aisle-3 for aisle in AISLES] + [aisle+3 for aisle in
AISLES]
667         #create_boarding_order_for_aisle_but_with_groups(window,middle,aisle,
window_seats)
668         # middle seats
669         middle_seats = [aisle-2 for aisle in AISLES] + [aisle+2 for aisle in
AISLES]
670         create_boarding_order_for_aisle_but_with_groups(middle,window,aisle,
middle_seats)
671         # aisle seats
672         aisle_seats = [aisle-1 for aisle in AISLES] + [aisle+1 for aisle in
AISLES]
673         create_boarding_order_for_aisle_but_with_groups(aisle,window,middle,
aisle_seats)
674
675         random.shuffle(window)
676         random.shuffle(middle)
677         random.shuffle(aisle)
678
679
680         window = [j for sub in window for j in sub]
681         middle = [j for sub in middle for j in sub]
682         aisle = [j for sub in aisle for j in sub]
683
684         boardingQueue1 = window+middle+aisle
685         for x in boardingQueue1:
686             if x not in boardingQueue:
687                 boardingQueue.append(x)
688
689         for passenger in boardingQueue:
690             passenger.append(assign_luggage())
691             passenger.append(0)
692
693         boardingQueue = cull_boarding_queue(boardingQueue)
694         test_cases.append(board_the_plane(boardingQueue))
695
696     print('By seat with groups: ', sum(test_cases)/len(test_cases))
697
698     return average(test_cases)
699
700 def prioritize_groups_boarding():
701
702     test_cases = []
703     for _ in range(N_TEST_CASES):
704         mainBoardingQueue = [[]]
705         priorityQueue=[]
706         boardingQueue=[]
```



```
763         mainBoardingQueue.append([])
764         current_boarding_section = 2
765     else:
766         priorityQueue.append([])
767         current_boarding_section = 1
768
769
770
771     random.shuffle(mainBoardingQueue)
772     random.shuffle(priorityQueue)
773     # flatten groups
774     boardingQueue = priorityQueue+['b']+mainBoardingQueue
775     boardingQueue = [j for sub in boardingQueue for j in sub]
776
777     #print(boardingQueue)
778
779     test_cases.append(board_the_plane(boardingQueue, True))
780
781     print('Priortizing groups: ', average(test_cases))
782
783
784
785
786
787
788
789
790
791
792
793
794
795
796
797
798
799
800
801
802 # modified steffen method
803 def steffen_modified_method():
804
805     test_cases = []
806     for _ in range(N_TEST_CASES):
807
808
809         rightOdd, leftOdd, rightEven, leftEven = [], [], [], []
810         steffenPerfected = [rightOdd, leftOdd, rightEven, leftEven]
811         # window seats
812         for row in range(1, NUM_ROWS+1):
813             for seat in range(-3, 4):
814                 #naughty boy
815                 if not is_not_disobedient() and seat != 0:
816                     steffenPerfected[random.randrange(0, 2)].append([row, seat+3,
assign_luggage(), 0])
817
818                 elif (seat > 0): #right side
819                     steffenPerfected[(row%2)*2].append([row, seat+3, assign_luggage
(), 0])
820
821                 elif (seat < 0): #left side
```

```
821         steffenPerfected[(row%2)*2+1].append([row, seat+3,
assign_luggage(),0])
822
823
824
825         random.shuffle(steffenPerfected[0])
826         random.shuffle(steffenPerfected[1])
827         random.shuffle(steffenPerfected[2])
828         random.shuffle(steffenPerfected[3])
829
830
831
832         steffenPerfected = [j for sub in steffenPerfected for j in sub]
833         test_cases.append(board_the_plane(steffenPerfected))
834
835     print('By steffen perefected: ', sum(test_cases)/len(test_cases))
836
837     return(average(test_cases))
838
839
840
841
842
843
844
845
846
847 plane = 'narrow body'
848
849 if plane == 'narrow body':
850     NUM_ROWS = 33
851     NUM_SEATS = 6
852     AISLES = [3]
853     F_SEC_START = 1
854     F_SEC_END = 11
855     M_SEC_START = 12
856     M_SEC_END = 22
857     A_SEC_START = 23
858     A_SEC_END = 33
859 elif plane == 'wide wing':
860     NUM_ROWS = 14
861     NUM_SEATS = 24
862     AISLES = [3,10,17,24]
863     F_SEC_START = 1
864     F_SEC_END = 5
865     M_SEC_START = 6
866     M_SEC_END = 9
867     A_SEC_START = 10
868     A_SEC_END = 14
869 elif plane == 'two entrance two aisle':
870     # simulating only the back half of the plane
871     NUM_ROWS = 20
872     NUM_SEATS = 7
873     AISLES = [2,6]
874     F_SEC_START = 1
875     F_SEC_END = 7
876     M_SEC_START = 8
877     M_SEC_END = 14
878     A_SEC_START = 15
879     A_SEC_END = 20
```

```
880 elif plane == 'two entrance two aisle first class':
881     # total loading time for 2E2A plane will be 2E2A + 2E2A first class
882     NUM_ROWS = 3
883     NUM_SEATS = 6
884     AISLES = [2,5]
885
886 # CHANGE THESE FOR SENSITIVITY
887 BAG_COEFFICIENT = (20,80,10)
888 NAUGHTY_BOY_COEFFICIENT = 0.18
889 COVID_CAPACITY = 0.5 #0, 0.3 0.5 or 0.7
890
891 N_TEST_CASES = 100
892 VISUALIZER = True
893 TIME_STEP = 0.1
894
895 # all measured in standard units (m,s,m/s etc)
896 AVERAGE_WALKING_SPEED = 0.8
897 AVERAGE_SEAT_PITCH = 0.78
898 TIME_TO_MOVE = AVERAGE_SEAT_PITCH / AVERAGE_WALKING_SPEED
899 FAMILY_TIME_TO_MOVE = 1.3 * TIME_TO_MOVE
900 NON_FAMILY_TIME_TO_MOVE = TIME_TO_MOVE
901 TIME_TO_SIT_OR_STAND = 2.5
902 TIME_TO_MOVE_PAST_SEAT = 2
903 # proportions of group sizes
904 SINGLE_PRINGLE_COEFFICIENT = 70
905 COUPLES_COEFFICIENT = 20
906 THREESOME_COEFFICIENT = 0
907
908 if VISUALIZER: im,fig = initialize_render()
909
910 #Data csv
911 import csv
912 fields = []
913 rows = []
914 index = []
915 #Add the indexing
916 for i in range(N_TEST_CASES):
917     index.append(i)
918
919 rows.append(index)
920
921
922
923 # BOARDING METHODS: comment out if not using
924 #random_boarding()
925 #section_boarding()
926 #seat_boarding()
927 #random_boarding_with_groups()
928 #section_boarding_with_groups()
929 #seat_boarding_with_groups()
930 #prioritize_groups_boarding()
931
932 # steffen methods can only be used with narrow body
933 #steffen_deeznuts()
934 #steffen_modified_method()
935
936 #naughty_boy_sensitivity()
937 #bag_sensitivity()
938
939 # field names add whatever field names that you are creating data for
```

```
940 fields = ['Index', 'Section']
```