# Final Report

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#### Intro

The included datasets come from two tables within the DSD relational database: Incarceration and Person. The incarceration table's unit of analysis is one booking into the jail and includes booking-related data such as the times into and out of the jail. The person table's unit of analysis is one person and includes demographic information like age and race. The "Person\_id" column is the common variable between them. (For this exercise, the real Person\_id has been suppressed and replaced with a unique random number to protect identities.)

The point of this exercise is to demonstrate how you think analytically as much as it is to arrive at the "correct" answers. Please provide your best answers to the questions below, using the tools and methods you deem most effective. Please submit written answers in a clear and concise form by the deadline. **Please also share your code so we can review it.** 

#### Exercise

2022.
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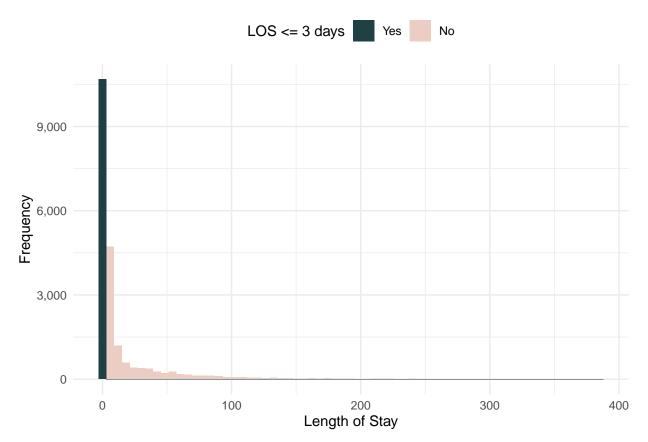
## 3. How many people were in the jail at the moment of the data extraction?

This would include not only the people who were booked during the analysis period, but the people who were booked before analysis period, but were not yet released. Based on this understanding of the question, I estimated approximately 16,456 people in the jail at the moment of the data extraction.

### 4. Consider the length of stay (LOS): the duration of each booking. Describe the LOS over the year analysis period. What insights (statistical and otherwise) does it provide you about variations in the jail population?

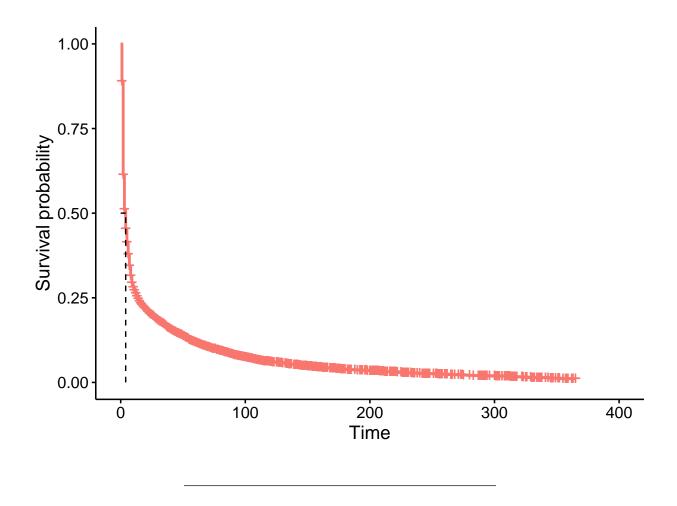
To look at length of stay during the analysis period, I think it is best to look at the bookings that were from July 1, 2021 to June 30, 2022. Bookings missing a release date were excluded from the visual review. Overall, the data seems to suggest that the people who are detained in jail tend to be changing frequently, if a little over half the jail population is only detained for 3 days or less.

- About 10,683 (51.3%) of bookings had a length of stay of at least 3 days.
- The most frequent length of stay being 2 days with a total of 6,041 (29.0%).
- There are about 25% of bookings that ranged from 10 days to 386.



From reviewing the distribution of the length of stay, it was determined that our vairable of interest was positively skewed. To determine what the median length of stay was for this population with 95% certainty, a survival analysis would make the most sense. In this scenario, we would be trying to understand the amount of time it takes a person who was booked to be released from jail during our analysis period.

From a survival analysis, the median release time for a person can be estimated to be 4 days, with a 95% CI [4,4]. One thing to note is that this median does differ from our review of the descriptive statistics because in this data set we kept bookings with a missing date where these bookings were previously excluded.



5. What was the average daily population in the jail during that year? Daily population ought to include anyone who spent even one minute in the jail in a given day. Please describe the methods/approach you used to answer this question. What tool did you use? What functions or other capabilities within the tool? (That is, help another analyst replicate what you did. Sharing code with your answers is encouraged but by no means required.)

Use a for loop to create a data set where if a booking occurred from Monday to Friday (5 days) the booking will be spread across 5 rows. This will create a data set where we can look at who was in jail each day of the analysis period. In order to correctly capture all the individuals in jail during this time, we will use the data set bookings\_during\_extraction created to answer question 3. The for loop I used can be found below.

```
# initialize table
booking_long = NULL

for(ith_booking in 1:nrow(bookings_during_extraction)){

# gather the current bookings information
booking = bookings_during_extraction[ith_booking,]

# if release date is missing, set this to the last day of the analysis period
# to capture each individual, even people who are still in jail after the analysis time constraint
if(!is.na(booking$release_out)) {
    release_out <- booking$release_out</pre>
```

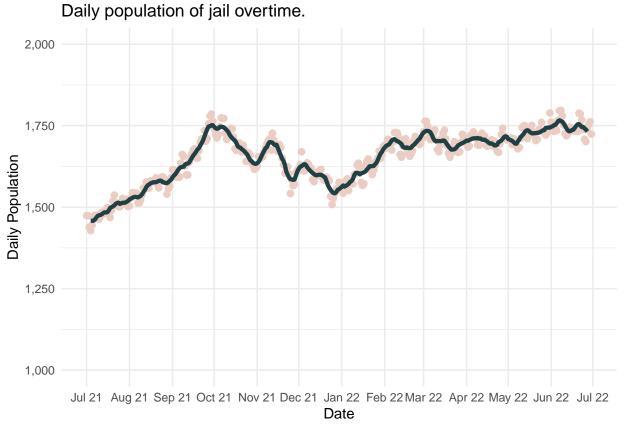
Now that we have a data set where a bookings length of stay is spread out by date, we can look at the daily population. We can do this by counting the number of bookings by date. This will provide us the daily population for our analysis period.

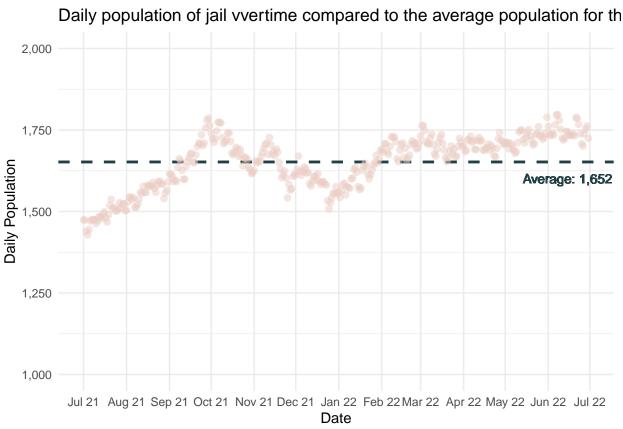
```
# count the number of people in jail each day
bookings_by_day <- booking_long %>%
  filter(date >= "2021-07-01" & date <= "2022-06-30") %>%
  arrange(date) %>%
  group_by(date) %>%
  count() %>%
  ungroup() %>%
  rename(daily_population = n)

# What is the average daily population for the analysis period?
(average_daily_population <- mean(bookings_by_day$daily_population))</pre>
```

#### ## [1] 1651.923

From the calculations above, we find that the average daily population in the jail during our analysis period is 1,652. By creating a data set in this way, we can also examine the trend of the daily population overtime. Examples of some graphics are shown below.





6. Which day during that year had the lowest daily population? What was the population that day? Which day had the highest daily population? What was the population that day?

July 04, 2021 was the day with the lowest daily population of 1,428. The day with the highest daily population was June 08, 2022 with a population of 1,797.

8.

- About 75.6% of people were only booked once