Emergency Homeless Shelter Availabilities Across Toronto*

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In this paper I analyze the occupancy and capacity of emergency homeless shelters across Toronto.

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^{*}Code and data are available at: https://github.com/cher-ning/shelter_occupancies

1 Introduction

Much like many other large cities around the world, Toronto is facing a homelessness crisis that seems to be ever-increasing. Whether it is because of environmental factors such as rising housing prices and lack of employment opportunities, or other issues such as drug abuse and mental health struggles, the homeless population of Toronto has risen to an alarming 10,811 in May 2023 (Draaisma 2023). The city council has recognized the urgency of this problem, declaring homelessness an emergency, and pledged their full efforts towards helping the homeless population move into safe homes (Draaisma 2023). Since then, the city of Toronto has focussed on securing more financial support from the provincial and federal government, requesting up to \$25 million to fund various housing projects. Notably, the Canada-Ontario Housing Benefit (COHB) aimed to move 1,600-2,000 households access and transition into permanent housing in 2023-2024 (Draaisma 2023).

In addition to these long-term plans that aim to move homeless individuals into stable homes, temporary shelters are also crucial during this transition process to provide short-term safety and resources. Research has shown that without access to shelter services, individuals can be pushed into substance-use relapses and hospitalizations, food insecurity, institutional circuitry through criminal justice systems, and other factors of instability (Kerman et al. 2024). Shelters also provide useful data to track and understand the current state of the city's battle against this crisis.

With that in mind, how did the city council's efforts turn out? Analyzing Toronto's shelter occupancy data from 2024, we do not see optimistic results. Average occupancy rate of emergency shelters per night remains at 99%, with no significant decrease across the months (ZipDataMaps, n.d.). Despite the evident high demand, there is even a decrease, albeit small, in shelters' funding capacity over the months (ZipDataMaps, n.d.).

In this paper, emergency shelter occupancy and availability will be compared across different months as well as different regions of Toronto to search for patterns or fluctutations in demand. The region of each shelter will be determined by its forward sorting area (FSA) code, or the first three digits of its postal code. As well, the regions that observe the highest frequency of failing to make efficient use of funding are identified as potential problem areas; further investigation into potential causes would be beneficial to understand if there are improvements that can be made so that the limited available funding can be maximized.

To achieve this, Section 2 will introduce the Toronto Shelter & Support Services dataset used for the present analysis and the data cleaning methods applied. The following Section 3 will then go more into depth to apply context to the patterns present in the data, aiming to understand the scope and urgency of the issues at hand.

2 Data

2.1 Overview

The dataset used for analysis is of shelter occupancy in the year 2024 (ZipDataMaps, n.d.), from Open Data Toronto (Gelfand 2022). The data is updated daily by the Toronto Shelter & Support Services and has the Open Data License (Section A.1). However, one limitation it faces is that it is unaudited, therefore there are limited verification methods to ensure that the shelter programs' records are accurate to their actual situation.

All shelters in this dataset are classified as either Emergency or Transitional programs, with Transitional locations providing more specialized programming and being exclusively offered to eligible individuals (ZipDataMaps, n.d.). With consideration for the greater accessibility of Emergency programs to the entire homeless population as well as the greater number of them shelters compared to Transitional, this analysis will focus on Emergency programs only (ZipDataMaps, n.d.). As well, all recorded shelters either measure capacity based on number of beds or rooms; similarly, this analysis will focus only on bed-based capacity shelters due to its greater prevalence (ZipDataMaps, n.d.). This means that all entries pertaining to non-Toronto based, Transitional type, or room-based capacity shelters were not considered in this analysis. It should also be noted that as this dataset contains only data from this year, only data from January 1 to September 26 was accessible as of writing this paper, and therefore only these entries were utilized in analysis.

With the remaining data entries, the variables of interest were the shelter's location, funding capacity, actual capacity, occupied beds, unoccupied beds, unavailable beds, and occupancy rate. Funding capacity refers to the maximum number of beds the location is able to offer, and represents the sum of actual capacity and unavailable beds. Unavailable beds refers to the number of spaces that are out of service due to logistical reasons (ZipDataMaps, n.d.). Actual capacity represents the number of beds that are usable per night, and should be the sum of occupied and unoccupied beds. After the dataset was cleaned for entries with empty values and invalid negative values, all entries in the remaining set were shown to adhere to these relations when tested.

The R Programming language (R Core Team 2023) was used for all data cleaning, testing, and analysis. The packages opendatatoronto (Gelfand 2022), tidyverse (Wickham et al. 2019), dplyr (Wickham et al. 2023), and readr (Wickham, Hester, and Bryan 2024) were used to simulate and download data. Packages tidyverse (Wickham et al. 2019) and janitor (Firke 2023) were then used to clean and test the downloaded dataset.

2.2 Results

This following section will use packages ggplot2 (Wickham 2016) and knitr (Xie 2024) to generate several plots and tables to assist with visualizing patterns in the dataset.

Table 1 provides a summary of the shelters' mean occupancy rates over the months. As observed, there is consistently high demand for shelters regardless of the time of year.

Table 1: 2024 Shelter Occupancy Rate Over the Months

Month	Occupancy Rate
Jan	98.82
Feb	99.17
Mar	98.53
Apr	98.94
May	99.13
Jun	98.77
Jul	99.12
Aug	99.37
Sep	99.36

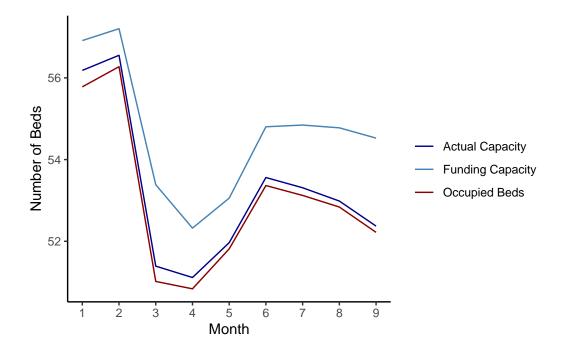


Figure 1: Shelter Capacity and Occupancy From Jan to Sep 2024

The occupancy rate can then be broken down into the mean funding capacity, actual capacity, and real number of occupied beds per month. The high occupancy rate observed in Table 1 makes it unsurprising to see the number of occupied beds consistently match so closely to the shelters' actual capacity in Figure 1. Notably, Figure 1 also shows a significant decrease in funding around March, which will be further discussed in Section 3. It is also important to note that despite the near 100% occupancy rate, there is often a gap between the funding capacity and actual capacity, suggesting that sometimes the given facilities may not be utilized at their maximum potential. This is a potential area of concern, and its implications will also be discussed in Section 3.

Another dimension to consider is the geographical region of the shelters. Still focusing on the mean number of occupied beds, Figure 2 shows a rather wide spread across different FSA regions, ranging from a minimum of 18.60 to a maximum of 87.74. Some of the regions reporting on the higher end include M5H, M6P, and M2H, which are neighbourhoods in Old Toronto, North York, and around High Park (ZipDataMaps 2024).

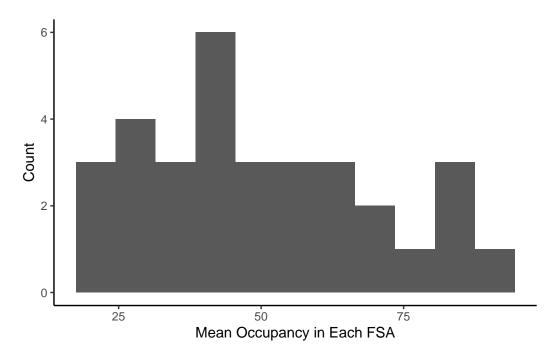


Figure 2: Average Bed Occupancy of Shelters in Each FSA Region

Another dimension to consider in analyzing which regions may be experiencing higher rates of homelessness is the number of distinct shelter locations made available there. Sorting for unique shelter ID's across the 32 FSA regions in this data set, Table 2 and Figure 3 show that regions most commonly have only 1-2 shelters each, but there are areas with particularly high density. Specifically, the region M5A located in Old Toronto (ZipDataMaps 2024) has 7 shelters, which is several times the number placed in other regions.

Table 2: Number of Emergency Shelter Locations Reported in Each Toronto FSA

FSA	Number of Shelter Locations
$\overline{\text{M5A}}$	7
M5R	4
M5C	3
M5S	3
M6H	3
M6K	3

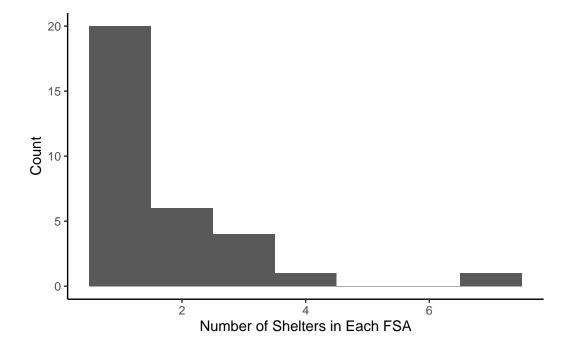


Figure 3: Number of Shelters in Each FSA

As shown in Figure 1, there is often a gap between a shelter's funding capacity and its actual capacity. This difference, reported as the number of unavailable beds in this dataset, represents resources that should have been available to the homeless population but was not on a particular night, either due to maintenance, renovations, outbreaks, or pest control issues (ZipDataMaps, n.d.).

Using similar methods as above to group data by FSA code finds that the average number of unavailable bed spaces per region is 1.30. Despite this promisingly low number, a closer look at the distribution in Table 3 reveals that there is a significantly higher average of unavailability in M2N and M5G, representing neighbourhoods in North York and Old Toronto respectively (ZipDataMaps 2024). The implications of this finding will be discussed further in Section 3.

Table 3: Number of Unavailable Beds Per Night in Different FSAs

FSA	Average Number of Unavailable Beds
$\overline{\text{M2N}}$	9.225
M5G	9.140
M6K	4.385
M5V	4.249
M4C	3.631

3 Discussion

3.1 Main Findings

Analysis of this dataset in Section 2.2 yielded several conclusions about the still-prevalent homelessness crisis here in Toronto. First, the consistently high occupancy rate throughout the year Table 1 suggests that the need for housing resources remains high, and is not particularly affected by time of year. However, as observed in Figure 1, there is a relatively steep dip in funding capacity around March that is not preceded by a noticeable dip in demand. This decrease is potentially due to the city emphasizing their efforts during winter months, as those are intuitively the most dangerous months for those living on the streets (Draaisma 2023). As a result, when winter ends and weather begins warming up in March, the additional funding is no longer available, causing this large and sudden decrease.

Moving onto FSA-based observations, it was found that there are several neighbourhoods including around Old Toronto and North York that have some of the highest concentration of emergency shelters and have highest funding capacities Table 2. Despite the high occupancy Figure 2 in these specific areas, it is important to remember that these findings are purely correlational and not causational. From this data alone, it is impossible to conclude whether these areas initially had greater rates of homelessness, thereby leading the government to spend more funds building shelters here, or that it is the higher concentration and availability of shelters here that will attract more homeless population to migrate here.

3.2 Implications

The observations made from this dataset reveal that the city council's effort have seen limited efficacy. Occupancy rates at emergency shelters remains alarmingly high all across Toronto with no sign of decreasing. Though these short-term solutions

3.3 Limitations and Next Steps

A Appendix

A.1 Data License Information

"Contains information licensed under the Open Government Licence – Toronto" (City of Toronto, n.d.). # Additional data details

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