

# Toronto Apartment Evaluations\*

Subtitle Here

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04 February 2022

## Abstract

First sentence. Second sentence. Third sentence. Fourth sentence.

## 1 Introduction

The city of Toronto introduced a new bylaw enforcement program in July 2017 that is used to ensure that the apartment building owners and operators comply with the necessary building maintenance standards, known as RentSafeTO. These standards are applied to all apartment buildings with 10 or more units or with 3 or more stories.

Owners of such apartments are required to register with RentSafeTO as well as maintain the standards defined by this program. Tenants need to contact their respective landlord in case they face an issue. These issues could either be vital, such as heat or hydro fault, or service requests, such as window flaws or common area cleaning. If the landlord does not comply to these requests, then according to the bylaws of RentSafeTO program, legal action can be taken against them.

Each property that falls under the program gets inspected by an officer and receives an evaluation score, which is made available to not only the landlord and tenant, but also to the potential tenants. If the score of a building is 86 or above, it will be evaluated again in the next three years. If it is between 66 and 85, it will be evaluated again within two years. If it is between 51 and 65, it will be evaluated again within a year. If a building gets a score of 50 or below, then the full building will undergo a comprehensive inspection.

Home safety evaluations are necessary to protect residents from potential hazards which may lead to personal injury if left unchecked. In a city like Toronto, where a lot of people do not own their personal property and live in rented apartments, it is critical that the landlord is kept informed of their building's condition so that they can ensure a tenant's safety. Dhekte hein ki kuch badiya sa milta hai ya nahi inn scores ke baare mein. Aayie shuru karte, bina kisi rukawat ke.

I think we should see which factors have the most effect on the evaluation score.

## 2 Data

```
package <- show_package("4ef82789-e038-44ef-a478-a8f3590c3eb1")
package
```

---

\*Code and data are available at: [www.github.com/ritvikpuri/sta304-paper-1](https://www.github.com/ritvikpuri/sta304-paper-1)

```
## # A tibble: 1 x 11
##   title      id      topics civic_issues publisher excerpt dataset_category
##   <chr>    <chr>    <chr> <chr>      <chr>    <chr>    <chr>
## 1 Apartment ~ 4ef82789-e~ <NA>  <NA>      <NA>      <NA>    <NA>
## # ... with 4 more variables: num_resources <int>, formats <chr>,
## #   refresh_rate <chr>, last_refreshed <date>
```

```
# get all resources for this package
```

```
resources <- list_package_resources("4ef82789-e038-44ef-a478-a8f3590c3eb1")
```

```
# identify datastore resources; by default, Toronto Open Data sets datastore resource format to CSV for
```

```
datastore_resources <- filter(resources, tolower(format) %in% c('csv', 'geojson'))
```

```
# load the first datastore resource as a sample
```

```
data <- filter(datastore_resources, row_number()==1) %>% get_resource()
```

```
sapply(data, class)
```

```
##           _id                                     RSN
##           "integer"                             "character"
##   YEAR_REGISTERED   YEAR_EVALUATED
##           "character"                           "character"
##   YEAR_BUILT       PROPERTY_TYPE
##           "character"                           "character"
##           WARD                                     WARDNAME
##           "character"                           "character"
##   SITE_ADDRESS     CONFIRMED_STOREYS
##           "character"                           "character"
##   CONFIRMED_UNITS  EVALUATION_COMPLETED_ON
##           "character"                           "character"
##           SCORE                                         RESULTS_OF_SCORE
##           "character"                           "character"
##   NO_OF_AREAS_EVALUATED  ENTRANCE_LOBBY
##           "character"                           "character"
##   ENTRANCE_DOORS_WINDOWS SECURITY
##           "character"                           "character"
##           STAIRWELLS    LAUNDRY_ROOMS
##           "character"                           "character"
##   INTERNAL_GUARDS_HANDRAILS  GARBAGE_CHUTE_ROOMS
##           "character"                           "character"
##   GARBAGE_BIN_STORAGE_AREA    ELEVATORS
##           "character"                           "character"
##   STORAGE_AREAS_LOCKERS  INTERIOR_WALL_CEILING_FLOOR
##           "character"                           "character"
##   INTERIOR_LIGHTING_LEVELS GRAFFITI
##           "character"                           "character"
##   EXTERIOR_CLADDING    EXTERIOR_GROUNDS
##           "character"                           "character"
##   EXTERIOR_WALKWAYS    BALCONY_GUARDS
##           "character"                           "character"
##   WATER_PEN_EXT_BLDG_ELEMENTS  PARKING_AREA
##           "character"                           "character"
##   OTHER_FACILITIES             GRID
##           "character"                           "character"
```

```
##           LATITUDE           LONGITUDE
##           "character"         "character"
##           X                   Y
##           "character"         "character"
```

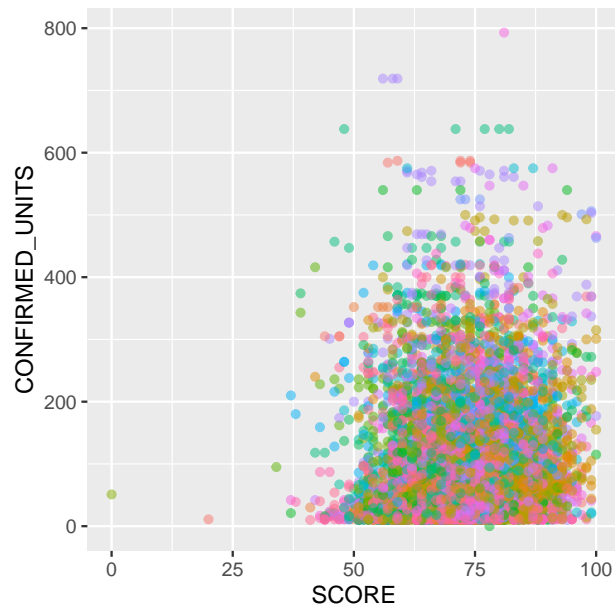
```
data[,1:5] <- sapply(data[,1:5], as.numeric)
data[,7:7] <- sapply(data[,7:7], as.numeric)
data[,10:11] <- sapply(data[,10:11], as.numeric)
data[,13:13] <- sapply(data[,13:13], as.numeric)
data[,14:35] <- sapply(data[,14:35], as.numeric)
```

```
## Warning in lapply(X = X, FUN = FUN, ...): NAs introduced by coercion
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```

```
view(data)
```

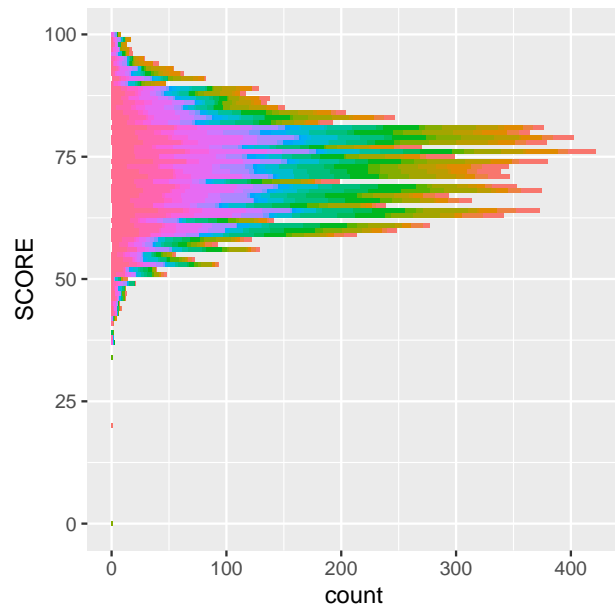
Talk more about it.

Also bills and their average (Figure ??). (Notice how you can change the height and width so they don't take the whole page?)



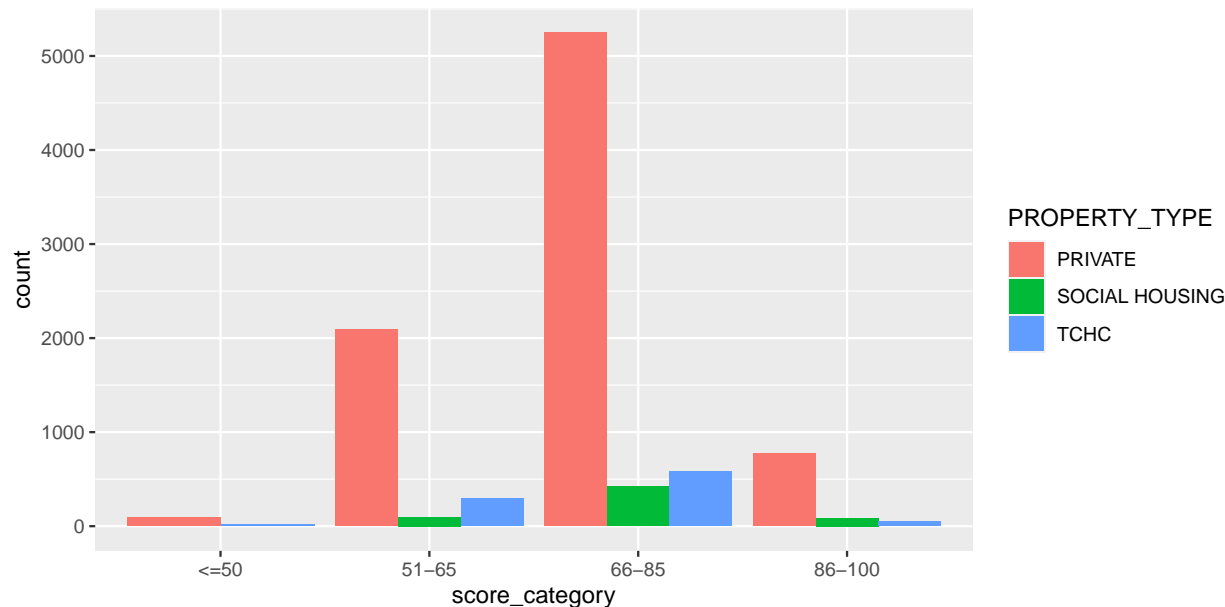
#### WARDNAME

- |                          |                        |
|--------------------------|------------------------|
| Beaches–East York        | Scarborough Southwest  |
| Davenport                | Scarborough–Agincourt  |
| Don Valley East          | Scarborough–Guildwood  |
| Don Valley North         | Scarborough–Rouge Park |
| Don Valley West          | Spadina–Fort York      |
| Eglinton–Lawrence        | Toronto Centre         |
| Etobicoke Centre         | Toronto–Danforth       |
| Etobicoke North          | Toronto–St. Paul's     |
| Etobicoke–Lakeshore      | University–Rosedale    |
| Humber River–Black Creek | Willowdale             |
| Parkdale–High Park       | York Centre            |
| Scarborough Centre       | York South–Weston      |
| Scarborough North        |                        |



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| Parkdale–High Park       | York Centre            |
| Scarborough Centre       | York South–Weston      |
| Scarborough North        |                        |



We can see how there are a lot of apartment building with a score that's between 66 and 85, which means they won't be evaluated for the next 2 years and are considered to be in good condition overall. But what makes the score of these buildings higher than the rest? And what do the few buildings with a score of more than 86 do that separates them from the rest?

### 3 Model

$$Pr(\theta|y) = \frac{Pr(y|\theta)Pr(\theta)}{Pr(y)} \quad (1)$$

Equation (1) seems useful, eh?

Here's a dumb example of how to use some references: In paper we run our analysis in **R** (R Core Team 2020). We also use the **tidyverse** which was written by Wickham et al. (2019) If we were interested in baseball data then Friendly et al. (2020) could be useful.

We can use maths by including latex between dollar signs, for instance  $\theta$ .

### 4 Results

### 5 Discussion

#### 5.1 First discussion point

If my paper were 10 pages, then should be at least 2.5 pages. The discussion is a chance to show off what you know and what you learnt from all this.

## **5.2 Second discussion point**

## **5.3 Third discussion point**

## **5.4 Weaknesses and next steps**

Weaknesses and next steps should also be included.

## Appendix

### A Additional details

## B References

<https://www.toronto.ca/community-people/housing-shelter/rental-housing-tenant-information/rental-housing-standards/apartment-building-standards/rentsafeto-for-tenants/>

<https://www.toronto.ca/community-people/housing-shelter/rental-housing-tenant-information/rental-housing-standards/apartment-building-standards/rentsafeto-for-building-owners/rentsafeto-building-evaluations-and-audits/>

Friendly, Michael, Chris Dalzell, Martin Monkman, and Dennis Murphy. 2020. *Lahman: Sean “Lahman” Baseball Database*. <https://CRAN.R-project.org/package=Lahman>.

R Core Team. 2020. *R: A Language and Environment for Statistical Computing*. Vienna, Austria: R Foundation for Statistical Computing. <https://www.R-project.org/>.

Wickham, Hadley, Mara Averick, Jennifer Bryan, Winston Chang, Lucy D’Agostino McGowan, Romain François, Garrett Grolemond, et al. 2019. “Welcome to the tidyverse.” *Journal of Open Source Software* 4 (43): 1686. <https://doi.org/10.21105/joss.01686>.