

THE ONE

Lin Han

August 9th, 2020



COVID – 19 Fatal Probability Calculator

-- Data Solution to Reveal Deeper

Reasons Behind COVID – 19 Fatal

Cases in Toronto

THE ONE

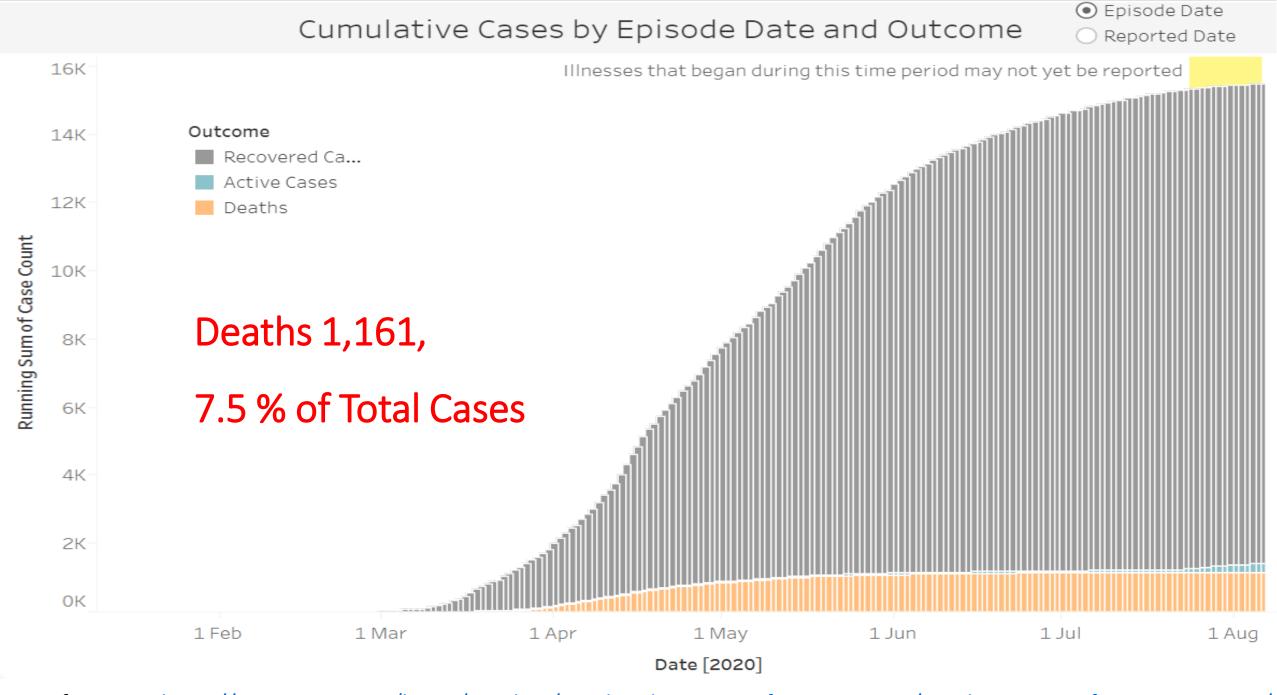
Lin Han - Schulich New Graduate



Agenda

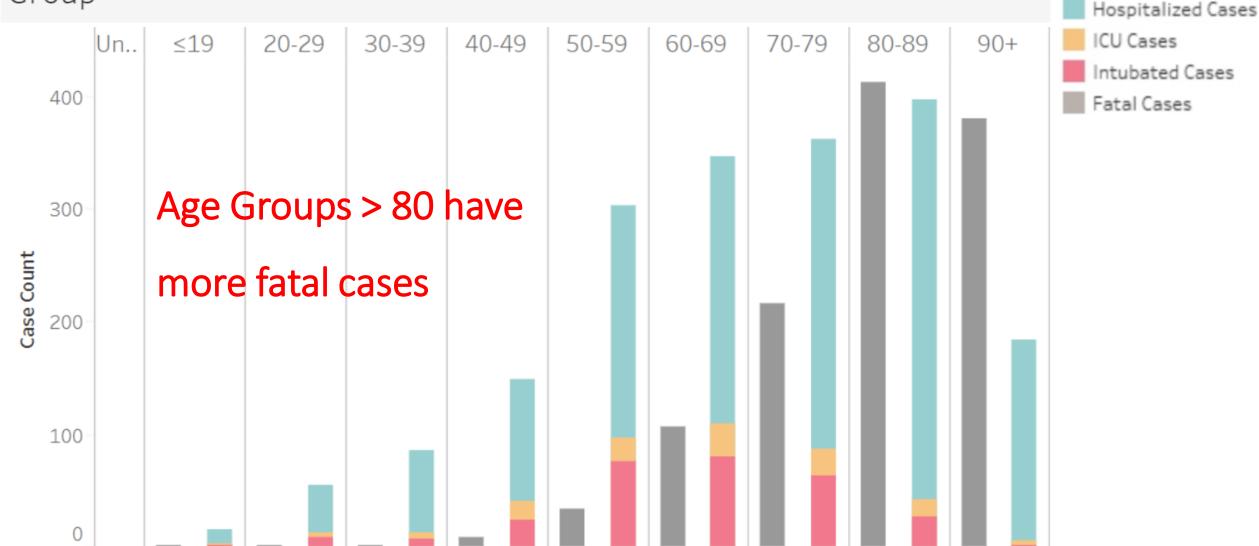
- I. Project Background
- II. COVID 19 Variables Analysis
- III. Fatal Probability Calculator
- IV. Calculation Examples
- V. Q&A



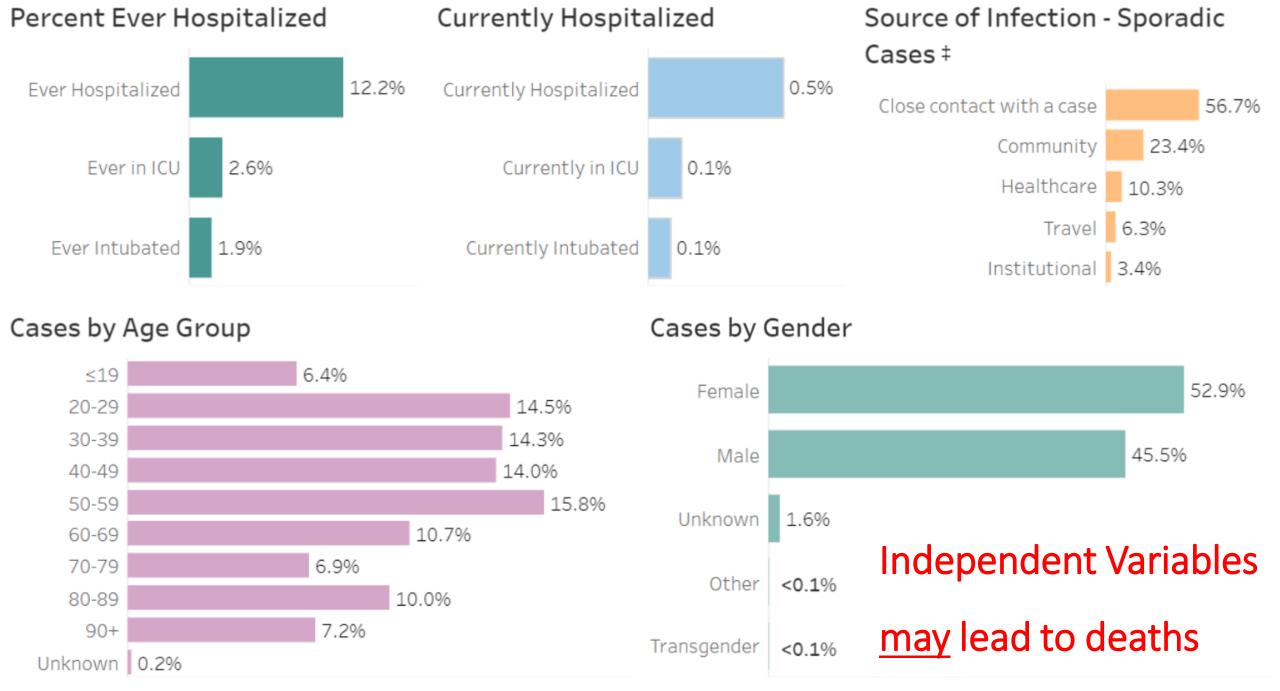


<u>City of Toronto: https://www.toronto.ca/home/covid-19/covid-19-latest-city-of-toronto-news/covid-19-status-of-cases-in-toronto/</u>

Number of COVID-19 Cases that Ever Resulted in Hospitalization, Intensive Care Unit (ICU) Admission, Intubation, and Deaths, by Age Group



<u>City of Toronto: https://www.toronto.ca/home/covid-19/covid-19-latest-city-of-toronto-news/covid-19-status-of-cases-in-toronto/</u>



<u>City of Toronto: https://www.toronto.ca/home/covid-19/covid-19-latest-city-of-toronto-news/covid-19-status-of-cases-in-toronto/</u>

Dig Deeper into Factors Cause Deaths

Removed 5 highly correlated variables:

'_id', 'currently_intubated', 'currently_in_icu', 'ever_in_icu', 'ever_intubated'.

	_id	outbreak_associated	confirmed	currently_hospitalized	currently_in_icu	currently_intubated	ever_hospitalized	ever_in_icu
_id	1	-0.26	-0.027	0.025	0.01	0.0071	-0.1	-0.048
outbreak_associated	-0.26	1	0.18	-0.0039	-0.014	-0.01	0.036	-0.062
confirmed	-0.027	0.18	1	0.017	0.011	0.0087	0.075	0.038
currently_hospitalized	0.025	-0.0039	0.017	1	0.52	0.42	0.2	0.19
currently_in_icu	0.01	-0.014	0.011	0.52	1	0.82	0.1	0.23
currently_intubated	0.0071	-0.01	0.0087	0.42	0.82	1	0.083	0.19
ever_hospitalized	-0.1	0.036	0.075	0.2	0.1	0.083	1	0.44
ever_in_icu	-0.048	-0.062	0.038	0.19	0.23	0.19	0.44	1
ever_intubated	-0.056	-0.058	0.033	0.19	0.24	0.23	0.37	0.81
duration	-0.0021	-0.15	-0.064	-0.014	-0.0053	-0.0025	-0.02	0.013
episode_month	0.71	-0.14	0.018	0.04	0.019	0.014	-0.1	-0.069

Toronto Open Data 'COVID-19 CASES IN TORONTO': https://open.toronto.ca/dataset/covid-19-cases-in-toronto/

Logistic Regression Model

-> 15 significant variables

'Age > 70' are more likely to die compared with 'Age < 50' because of COVID – 19.

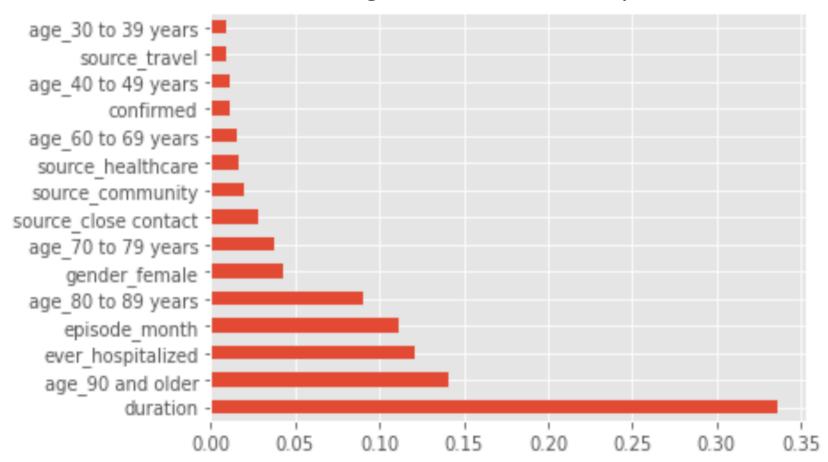
Results: Logit

===========	=======	:======	=======	======	:======	
Model: Dependent Variable: Date: No. Observations: Df Model: Df Residuals: Converged: No. Iterations:	Logit outco 2020- 10364 14 10349 1.000	AIC 39 BIC Log LL- LLR	Pseudo R-squared: AIC: BIC: Log-Likelihood: LL-Null: LLR p-value: Scale:			
	Coef.	Std.Err.	z	P> z	[0.025	0.975]
confirmed	-1.1483	0.2032	-5.6516	0.0000	-1.5465	-0.7501
ever_hospitalized	1.7442			0.0000		
duration	-0.0285			0.0001		
episode_month	-0.6323		-14.6424			
age_30 to 39 years	-2.8325	1.0115	-2.8004	0.0051	-4.8150	-0.8501
age_40 to 49 years	-1.2285	0.4738	-2.5926	0.0095	-2.1572	-0.2998
age_60 to 69 years	1.2970	0.2041	6.3538	0.0000	0.8969	1.6970
age_70 to 79 years	2.4162	0.1890	12.7846	0.0000	2.0458	2.7866
age_80 to 89 years	2.8525	0.1819	15.6858	0.0000	2.4961	3.2089
age_90 and older	3.5033	0.1886	18.5792	0.0000	3.1337	3.8728
source_close contact	-1.0371	0.1651	-6.2815	0.0000	-1.3607	-0.7135
source_community	-0.8230	0.1910	-4.3077	0.0000	-1.1974	-0.4485
source_healthcare	-0.6684	0.2265	-2.9514	0.0032	-1.1123	-0.2245
source_travel	-1.4308	0.3209	-4.4593	0.0000	-2.0597	-0.8019
<pre>gender_female</pre>	-0.5194	0.0906	-5.7314	0.0000	-0.6970	-0.3418
			•			

Variables Importance Rating

Top variables correlated to deaths: 'Duration of episode_date & reported_date',

'Age 80 and older', 'Episode month'







COVID – 19 Fatal Probability (P)

```
= 1 / 1 + \exp{-[confirmed *(-1.148) + ever hospitalized * (1.744) + duration * (1.744) + dura
 (-0.029) + episode month * (-0.632) + age 30 to 39 years * (-2.833) + age 40 to 49
years * (-1.229) + age 60 to 69 years * (1.297) + age 70 to 79 years * (2.416) +
age 80 to 89 years * (2.852) + age 90 and older * (3.503) + source close contact *
 (-1.037) + source community * (-0.823) + source healthcare * (-0.668) +
source travel * (-1.431) + gender female * (-0.519)]}
```

(Except for variables 'duration' and 'episode_month' are integer data type, other variables are dummy variables: 0 or 1.)

Fatal Probability Calculation – Example 1

Given a confirmed COVID – 19 patient A, who has in hospital, and duration is 10 days, episode month is 7, age 35, and source is community, gender female.

COVID – 19 Fatal Probability (P)

```
= 1 / 1 + \exp\{-[1 * (-1.148) + 1 * (1.744) + 10 * (-0.029) + 7 * (-0.632) + 1 * (-2.833) + 0 * (-1.229) + 0 * (1.297) + 0 * (2.416) + 0 * (2.852) + 0 * (3.503) + 0 * (-1.037) + 1 * (-0.823) + 0 * (-0.668) + 0 * (-1.431) + 1 * (-0.519)]\}
= 1 / 1 + \exp[-(-8.293)] = 1 / 3996.80 = 0.025 \% -> will not likely die
```

Fatal Probability Calculation – Example 2

Given a confirmed COVID – 19 patient B, who has in hospital, and duration is 30 days, episode month is 2, age 90, and source is close contact, gender male.

COVID – 19 Fatal Probability (P)

```
= 1 / 1 + \exp\{-[1 * (-1.148) + 1 * (1.744) + 30 * (-0.029) + 2 * (-0.632) + 0 * (-2.833) + 0 * (-1.229) + 0 * (1.297) + 0 * (2.416) + 0 * (2.852) + 1 * (3.503) + 1 * (-1.037) + 0 * (-0.823) + 0 * (-0.668) + 0 * (-1.431) + 0 * (-0.519)]\}
= 1 / 1 + \exp[-(0.928)] = 1 / 1.3953 = 71.67 \% -> will likely die
```



Reference

1. Picture source:

http://www.pentictonherald.ca/news/national news/article a29f7353-8eb4-5fd5-a7e5-5d31d16d43b5.html

2. City of Toronto, COVID-19: Status of Cases in Toronto:

https://www.toronto.ca/home/covid-19/covid-19-latest-city-of-toronto-news/covid-19-status-of-cases-in-toronto/

3. Toronto Open Data 'COVID-19 CASES IN TORONTO':

https://open.toronto.ca/dataset/covid-19-cases-in-toronto/

4. Probability Calculation Using Logistic Regression

https://docs.tibco.com/pub/sfire-dsc/6.5.0/doc/html/TIB_sfire-dsc_user-guide/GUID-C4D05ED0-3392-4407-B62A-7D29B26DC566.html