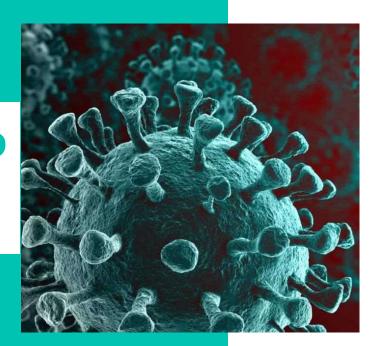
### **Text Mining on COVID-19**Patients' Data

**Data Science Capstone Project** 

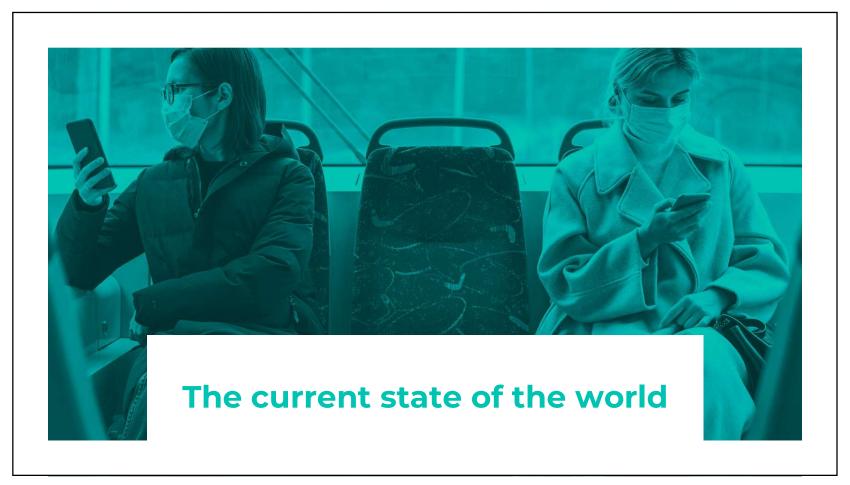


**Princess Allotey** 

O2. METHODS How?  O3. RESULTS What?  O4. DISCUSSION What? & What next?  O5. REFERENCES			
How?  O3. RESULTS What?  O4. DISCUSSION	05.	REFERENCES	
How?  O3. RESULTS	04.		
How?	03.		
O2. METHODS	07		
	02.	METHODS	
O1. INTRODUCTION What & Why?	01.		

# 01. INTRODUCTION

How have your day-to-day activities been affected by COVID-19?\*



### **TIMELINE**

- December 2019 discovered in Wuhan, China
- March 11<sup>th</sup>, 2020 declared a pandemic
- September 14<sup>th</sup>, 2020 29 million confirmed cases worldwide, and close to 1 million deaths



COVID-1

### PROJECT GOAL & VISION OF THE FUTURE

Predicting the outcome of future patients from electronic health records



# **MAIN SIGNIFICANCE** Triage practices



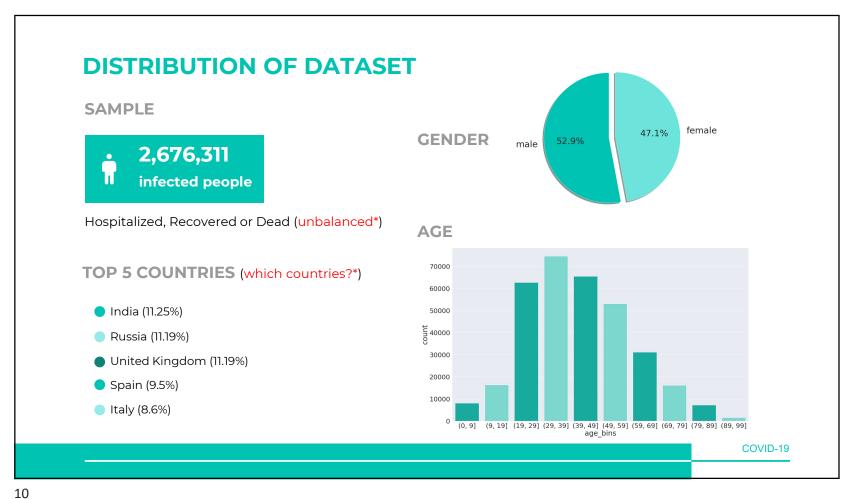
### **DATA SOURCE**

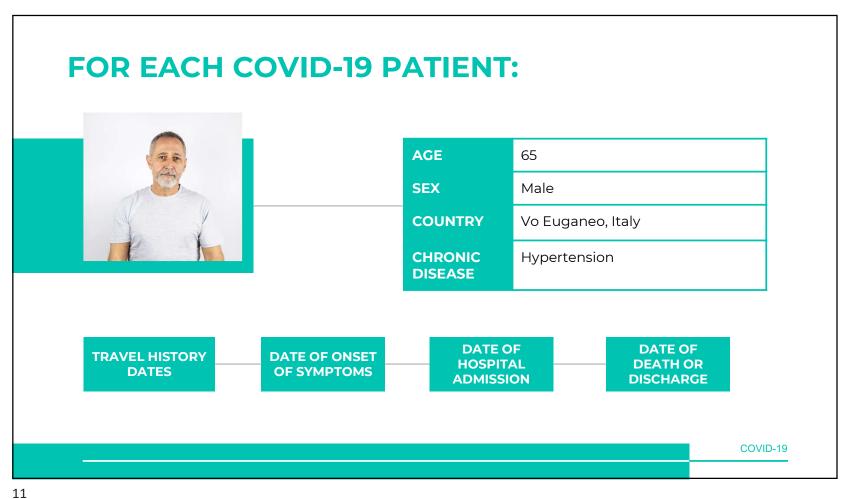


Institute for Health Metrics and Evaluation

University of Washington

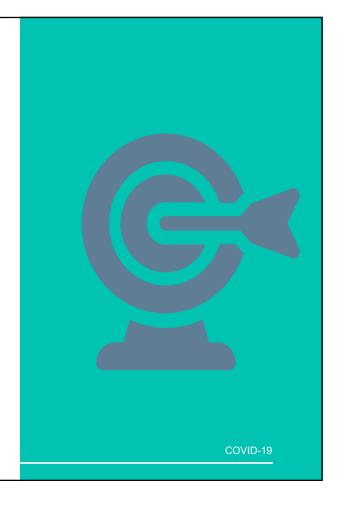
COVID-19





### **COLUMNS OF INTEREST**

- symptoms
- outcome
- additional information
- notes for discussion
- chronic disease
- travel history location

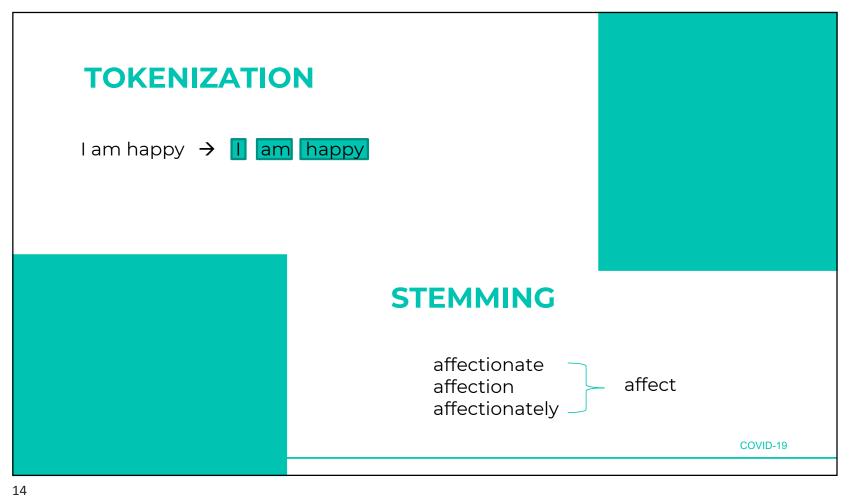


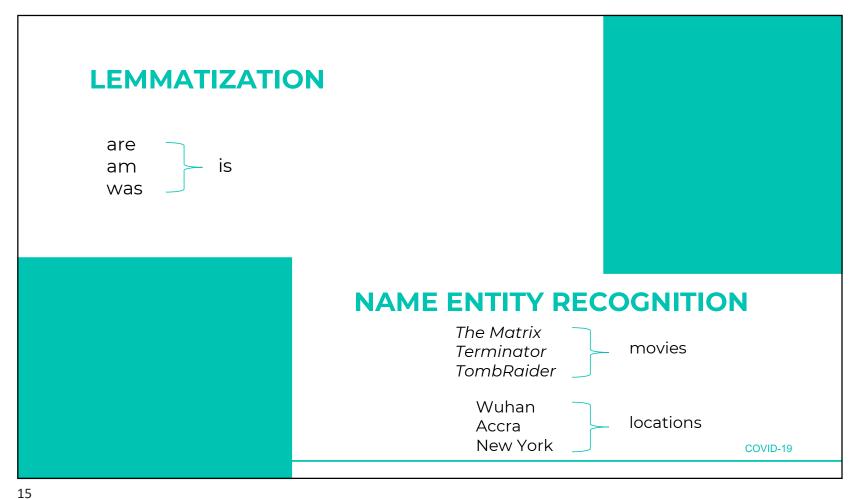
### **MODEL 1**

- Model 1: Natural Language Processing (NLP)
- Reason: Used in Electronic Health Records (Medical Informatics)

- Assumption: Text data is generated from the COVID patient
- Tool: Natural Language Toolkit (NLTK)
  - unique functions
  - popular toolkit

COVID-19





### **Text mining process:**

- Punctuation
- Stemming
- Lemmatization
- Tokenization

### Text analysis process & model training:

- 5 machine learning models
- Voting Classifier (Ensemble Method)

## CORE NLTK FUNCTION

COVID-1

### **NLP MODEL BUILT FOR:**

#### **TEXT MINING**

The outcome of a COVID patient dependent on:

- travel history location\*
- symptoms\*
- chronic disease\*
- additional information

#### **ANALYSING DATES**

Predicting length of time in:

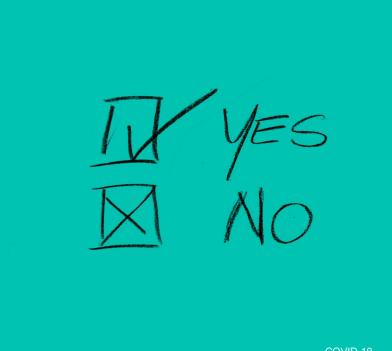
- hospital after onset of symptoms given additional information on patient
- hospital after hospital admission given symptoms

Why do you think these two are different?

COVID-19

### MODEL 2

- **Model 2:** Logistic Regression
- **Reason:** Integrates previous explorations
- Predicts the likelihood that a patient will recover



COVID-19

**Y:** 

Outcome

X:

- Age
- Sex
- Chronic disease binary
- 2 date differences
- Symptoms

TARGET AND FEATURES

COVID-19



• Ensemble Method Accuracy: 85.19%

• F1-score:

o Dead: 0.47

Recovered: 0.91

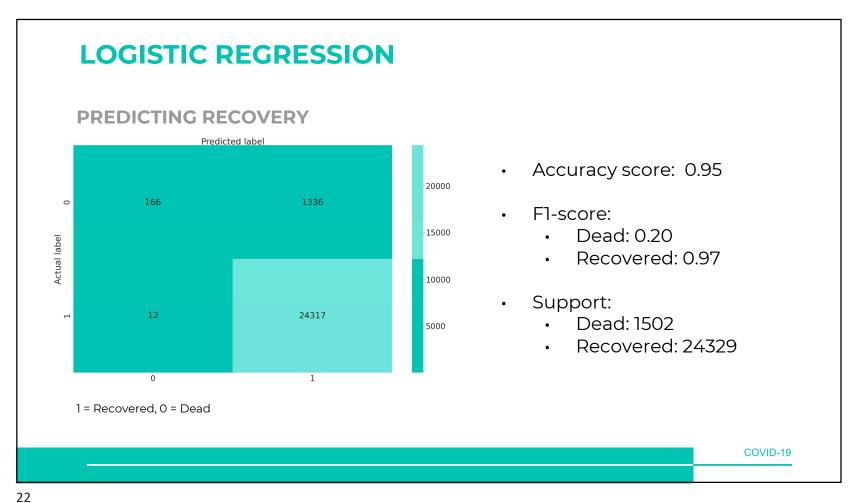
• Support:

o Dead: 259

Recovered: 1017

NLP: OUTCOME OF A COVID PATIENT DEPENDENT ON ADDITIONAL INFORMATION

COVID-19



# 04. DISCUSSION

What did I find?

What are my next steps?

1. Conclusion: Using text and dates data, I prepared Logistic Regression and Natural Language Processing models to determine the likelihood that a patient will recover from COVID-19

#### 2. Future Work:

- Explore and learn from similar case studies
- Consider a more balanced dataset (recovered and dead)\*

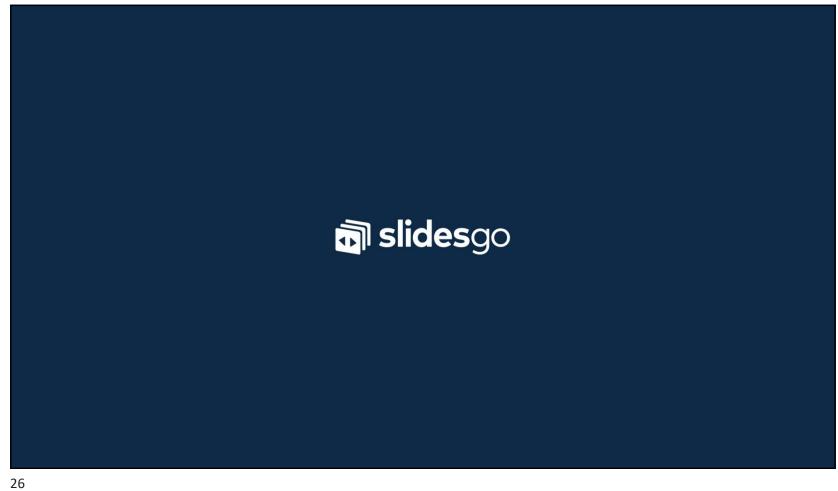
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### 05. REFERENCES

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### Let's tackle COVID-19 quickly and efficiently!

Do you have any questions? princess.allotey@centre.edu +1 859 319 0168

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