# Phase 3 Project

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## Overview

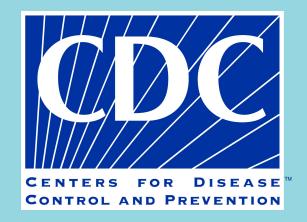
Client: Providence and Medical Centers



Source: CDC

**Target Variable: Seasonal Vaccines** 

**Goal**: **Minimize** vaccine wastage when ordering vaccines.



## **Business Problem**

#### Vaccines are wasted due to:

- **→** Expire date
- → Supply Chain Issue



## **Real Life Example:**

> **1.1 billion** Covid Vaccines were estimated to be **wasted** due to expired vaccines and supply chain issues.

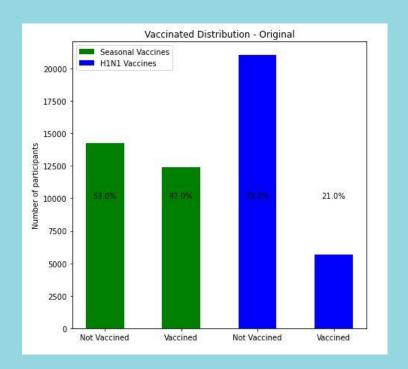
# **Data Understanding**

# Number of observation: **26,000** participants

- → 36 different survey question
- → Roughly **50%** taken seasonal vaccine

Target Variable: <u>Seasonal</u>

Flu Vaccine



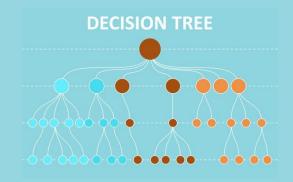
## Method

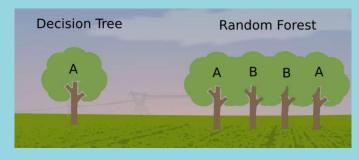
# **Preprocessing Steps**

- > Split Dataset to **Training** and **Testing** Sets
- > Prepare Datasets
  - (Missing values, Scalar, Dummy Variables)
- Fit Model with Training Dataset
- > Make Predictions

## Models

- > Simple Baseline Model
  - Decision Tree
- Complex Model
  - Random Forest
- > Tuned Model
  - Random Forest with Tuned Hyperparameters





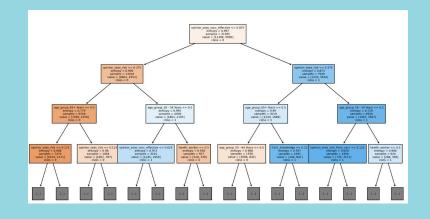
# Simple Decision Tree

#### **Decision Tree Classification:**

> Split dataset based on features in order to reduce entropy.

**Parameters**: Default, criterion = entropy





# **Decision Tree - Analysis**

Perfect Training
Accuracy
100%

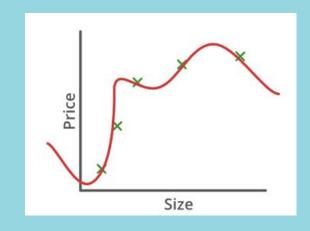


Low Testing Accuracy 68.21%



Overfitting with Training
Dataset
(Greedy Algorithm)

Overfitting - A model trained to perfectly predict the training dataset



# Decision Tree → Random Forest

<b>Decision Tree</b>	Random Forest		
+ Interpretable	+ Resilient to <b>overfitting</b>		
- <b>Prone</b> to overfitting	+ Resistance to <b>noise</b>		
	- <b>Long</b> Computational Time		

## Random Forests

#### **Random Forest Classification:**

Create multiple Decision Trees with different set of features.

#### **Parameters:**

Max Depth = 10, Max Features = None, criterion = entropy



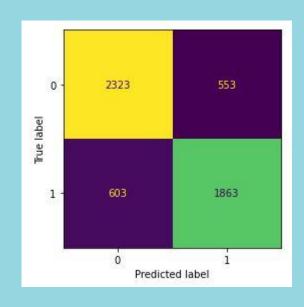
Training Accuracy: 83.45%

Testing Accuracy: 77.89%

# Random Forest with Tuned Hyperparameters

## **Tuning with Grid Search**

- Compares models with different parameters
- Over 7,000 combinations





**Training Accuracy: 92.59%** 

**Testing Accuracy: 78.36%** 

# Model Comparison

Model	Training Accuracy	Testing Accuracy	Recall Score
Decision Tree	100%	68.21%	65.33%
Random Forest	83.45%	77.89%	75.55%
Random Forest with hypertuning	92.59%	<u>78.36%</u>	<u>75.55%</u>

# Real Life Application (Example)

#### Data Taken from Testing Dataset

Number of Patients	Predicted Taken	Patients Actually Taken	Number of Vaccines + 5% *	Vaccines Wasted
5342	2416	2466	2416 + 121 = 2537	71

<sup>\*</sup> Supply should be <u>slightly more</u> in case of faulty or new patients Subject to change based on location. (Major Cities, Small Towns)

## Recommendation

Most Important Features:

Opinion\_Seasonal\_Risk, Opinion\_Seasonal Vaccine\_Effective, Doctor\_Recommendation\_Seasonal

Question Concerning Vaccines → **More Likely** to take vaccine

#### Recommendation:

- Only include question regarding current vaccine focus
- Occupation have little to no influence in prediction

### Conclusion

- > Random Forest Performs **Better** Than Decision Trees
- > 92.59% Training Accuracy
- > 78.36% Testing Accuracy

- Use Model to obtain rough estimate number of vaccines
- > 71 wasted vaccines vs 2,876 \*

\* Assuming everyone needs a vaccines

# Next Steps

- 1. More Tuning
- 2. Different Classification Models
  - a. KNN
  - b. XGBoost

# 3. Better Survey Questions

- a. Demographic
- b. Religion

# Thank You...Questions?

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**GitHub:** https://github.com/Tommyphung1/phase\_3\_project