# Pattern Recognition and Machine Learning (11482) Final Report

Using Machine Learning to Predict Diabetes Terence Lam (u3206488) Uyen Nguyen(u3206201)

# **Project and Problem Summary**

- Diabetes is a common disease in nowadays society
- As in 2019, Approximately 1.5 million people die to diabetes (WHO 2021)
- Objective: use machine learning to diagnose and predict existing or potential diabetes disease on a patient

### **Dataset**

- Pima Indians Diabetes Database
- From the National Institute of Diabetes and Kidney Digestive
- 786 instances
- 9 attributes (features): pregnancies, glucose, blood pressure, skin thickness, insulin, BMI, diabetes pedigree function, age, and outcome (label 0, 1)
- Clean up the dataset

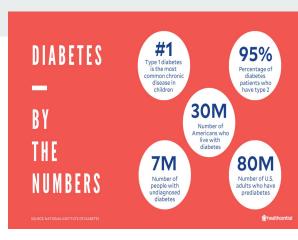


Figure 1: Diabetes by the numbers (WriterMarch 10 et al., n.d.)

# Methodology

- 1. Input the dataset
- 2. Pre-processing
- 3. EDA
- 4. Scale the dataset
- 5. Experiment the training dataset with models
- 6. Evaluate the score with evaluation strategies

### **Evaluation Strategies:**

- Accuracy
- ROC curve AUC
- F1-score
- 7. Compare and analyse the results

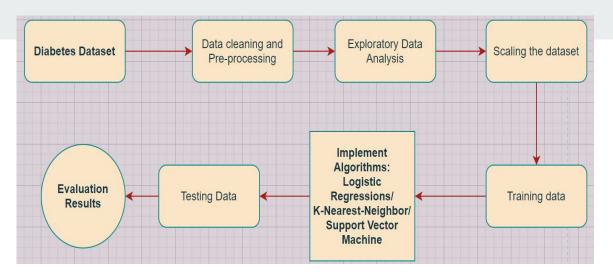


Figure 2: Methodology and procedure of the project

### **Pre-processing**

# **Exploratory Data Analysis**

### The modification process:

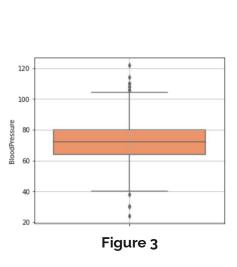
Relationship interpretation:

- Depending on the values distribution of each variables:

The person who has high level of BMI,
Glucose, Skinthickness can get diabetic.

- Replace "0" values with mean in Glucose, Blood Pressure, Skin Thickness, BMI.
- Replace "0" values with median as for Insulin.
- Cut down the variable named Diabetes Pedigree Functions

- Getting rid of the outliners



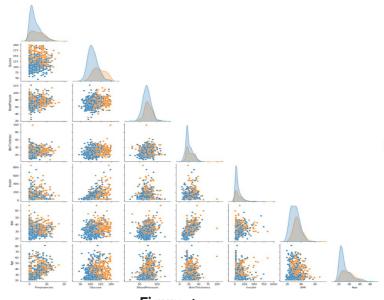




Figure 4

Figure 5

### **Processing the Dataset**

- Standard Scale the dataset to standardize the variables
- Processing the data with 80% training and 20% testing

# Models implementation and Hyperparameters tuning

### **Classification report**

Accuracy

Precision

Recall

F1-score

**AUC** 

Logistic Regression	C=10		
K-Nearest-Neighbor	{'metric': 'euclidean', 'n_neighbors': 25}		
Support Vector Machine	{'C': 100, 'gamma': 0.001, 'kernel 'rbf'}		

Figure 6

Model	Accuracy	F1-score	AUC-score
Logistic Regression	0.7237	0.72	0.84
KNN	0.7960	0.80	0.79
SVM	0.7829	0.78	0.82
Tuning LR	0.7649	0.80	0.7397
Tuning KNN	0.7549	0.75	0.6857
Tuning SVM	0.7516	0.80	0.7447

### Contribution and future works

### Contributions

- People can improve their health by applying their variables statistic on the predictive models
- Helpful in future research in the medical field

### Future works

- Lack of time and experience
- Experiment on other learning models and evaluation strategies
- Try different variations of learning models and evaluation strategies used in this project e.g. time-dependent ROC curve, amalgam KNN
- Experiment on different datasets

# Thank you for listening!

### **Teamwork Contributions**

Uyen: Data processing and model learning.

Terence: Data cleaning and formatting.

The whole team: Powerpoint and report preparation.

### Reference

https://www.healthcentral.com/condition/diabetes

https://towardsdatascience.com/how-and-why-to-standardize-your-data-996926c2c832