

Machine Learning Project

# **DIABETES PREDICTION SYSTEM**

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# Plan

**Introduction**

III

**Performance Evaluation**

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II **Different Algorithms**

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# Introduction

Technology and Health Care Services



# I Related work

## The Paper :

# Machine Learning-Based Unified Framework for Diabetes Prediction

## The Dataset :

### PIMA

The Prima Indian Diabetes Dataset has been used in this study, provided by the UCI Machine Learning Repository. The dataset has been originally collected from the National Institute of Diabetes and Digestive and Kidney Diseases.

## The Features:

Pregnancies

Glucose

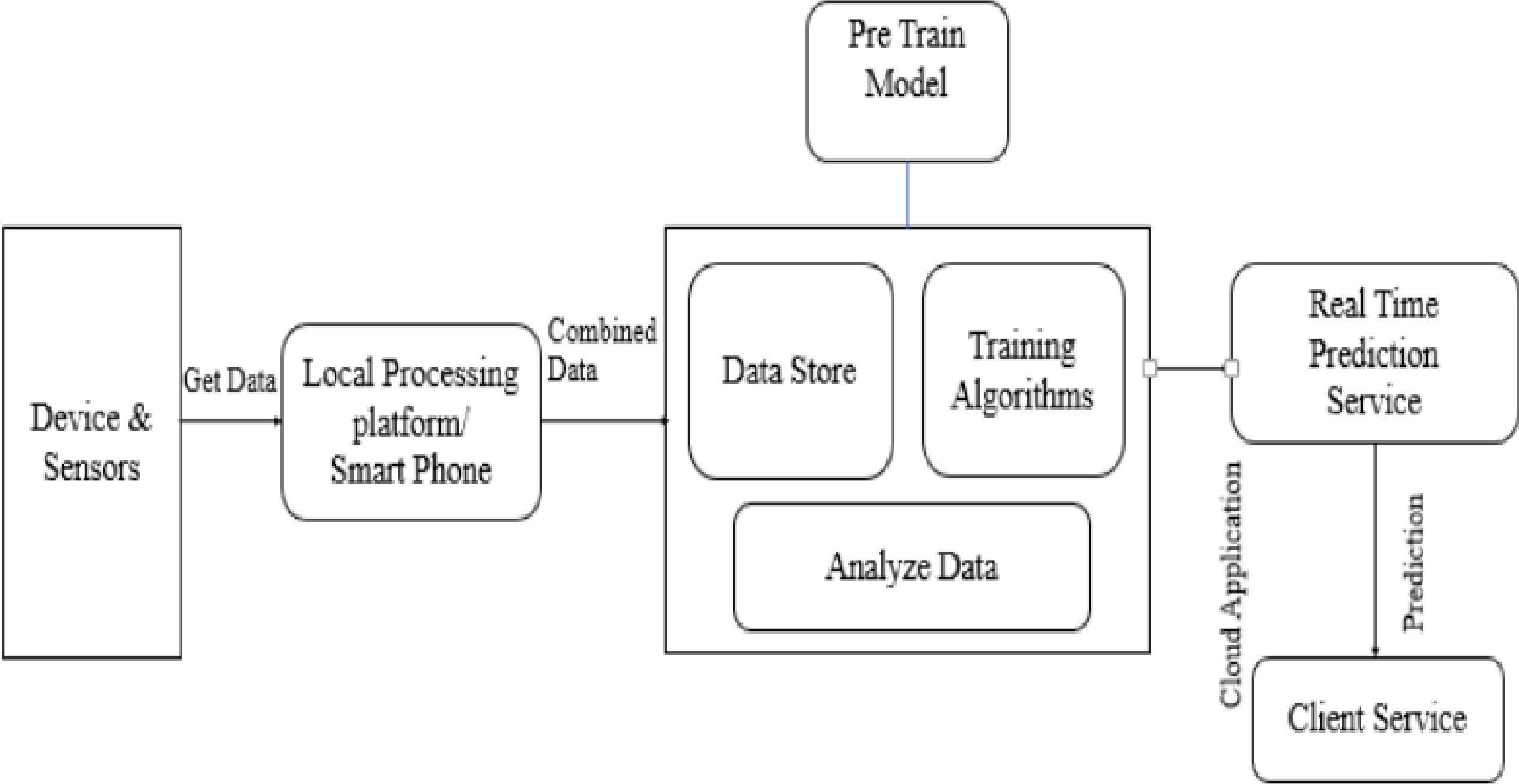
Blood Pressure

Skin Thickness

Insulin

BMI

# The Project Framework:



# II Classification Algorithms



# The Classification Algorithms:

- Artificial Neural Network (ANN)
- Support Vector Machine (SVM)
- Logistics Regression (LR)
- Decision Tree (DT)
- Random Forest (RF)
- Naive Bayes (NB)



# Performance Evaluation

# The Classification Performance Measurements

- Different **statistical measurement aspects** such as **accuracy, precision, F-1**
- These classification measurement factors are calculated by the terms:  
True Positive (**TP**), False Positive (**FP**), True Negative (**TN**), and False Negative (**FN**)

**True Positive (TP):** Prediction results are yes and the patient has diabetes.

**True Negative (TN):** Prediction results are no and the patient does not have diabetes.

**False Positive (FP):** Prediction results are yes but the patient does not actually have the diabetes.

**False Negative (FN):** Prediction results are no but the patient has diabetes.

$$\text{Accuracy} = (TP+TN)/(TP+FP+FN+TN)$$

$$\text{Recall} = TP/(TP+FN)$$

$$\text{Precision} = TP/(TP+FP)$$

$$\text{F-1} = 2 * (\text{Recall} * \text{Precision})/(\text{Recall} + \text{Precision})$$

# The Classification Performance Measurements

Mesurements Techniques	NB	RF	LR	ANN	SVM	DT
Accuuracy	0.73	0.74	0.71	0.70	0.74	0.71
Precision	0.61	0.62	0.57	0.58	0.63	0.58
F-1	0.63	0.63	0.62	0.54	0.60	0.61

=>so the applied Algorithm is : **SVM**

# IV Project Realisation

## Tools :



**Anaconda**



**Jupyter Notebook**



**Python**



**Sci Hub**



**Pycharm**

# IV ITS'D DEMO

# Code Snippet of the implementation in our code

```
19 def result(request):
20     #getting the data
21     data_frame = pd.read_csv(r"C:\Users\Lina Ben Salem\Desktop\diabetes project\diabetes.csv")
22     #splitting the data
23     feature_col_names = ['Pregnancies', 'Glucose', 'BloodPressure', 'SkinThickness', 'Insulin', 'BMI',
24                          'DiabetesPedigreeFunction', 'Age']
25     predicted_class_names = ['Outcome']
26
27     X = data_frame[feature_col_names].values # predictor feature columns (8 X m)
28     y = data_frame[predicted_class_names].values # predicted class (1=true, 0=false) column (1 X m)
29     split_test_size = 0.30
30
31     X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=split_test_size, random_state=42)
32     # test_size = 0.3 is 30%, 42 is the answer to everything
33
34
35     svm_model = SVC(kernel='linear', C=1, random_state=42)
36
37     svm_model.fit(X_train, y_train.ravel())
38
39     val1 = float(request.GET['n1'])
40     val2 = float(request.GET['n2'])
41     val3 = float(request.GET['n3'])
42     val4 = float(request.GET['n4'])
43     val5 = float(request.GET['n5'])
44     val6 = float(request.GET['n6'])
45     val7 = float(request.GET['n7'])
46     val8 = float(request.GET['n8'])
47
48     pred = svm_model.predict([[val1, val2, val3, val4, val5, val6, val7, val8]])
49     result1 = ""
50     if pred == [1]:
51         result1 = "Positive"
52     else:
53         result1 = "Negative"
54
```



# Conclusion

# Conclusion & Prespective

- **Xgboost**
- **Lazy Predict**

Thank you  
for your  
Attention