



## TEAM MEMBERS

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# PREDICTING BREAST CANCER

**with Machine Learning**

<https://github.com/MajoGarciaMontes/FINAL-PROJECT>

# CONTENT

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**01**

Objective / main  
questions

**02**

Architecture of our ML  
solution

**03**

Data gathering and  
cleansing

**04**

ML model design

**05**

API integration

**06**

Visualization /  
Dashboard

**07**

Conclusions / Next  
steps

# OBJECTIVE / MAIN QUESTIONS

## CONTEXT

- *'A woman born today has about a 1 in 8 chance of being diagnosed with breast cancer at some time during her life'*  
*National Cancer Institute*
- *The costs of having more advanced tests done to determine whether a tumor is benign or malignant can go up exponentially, so not every patient will be able to have them done.*



## OBJECTIVE

**Develop a Machine learning (ML) model and train it with relevant breast cancer variables to predict the probability of developing breast cancer.**

## MAIN QUESTIONS TO SOLVE



- **What classification fits the patient? (benign / malignant tumor).**
- **Which is the best model to use?**
- **What is the accuracy of the classification?**

# ARCHITECTURE OF OUR ML SOLUTION

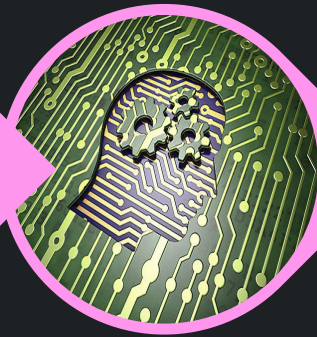
During the last 6 months we have learned to create data solutions:



**PROJECT  
PROPOSAL**



**ETL**



**ML  
DESIGN**



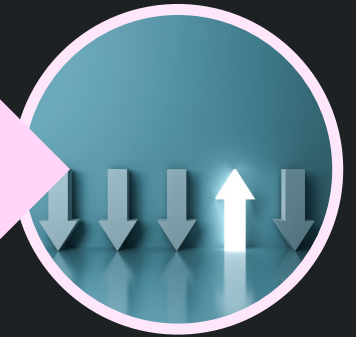
**API**



**DASHBOARD  
CODING**



**USER  
TESTING**



**PROJECT  
EVALUATION**

# DATA GATHERING / CLEANSING

## GATHERING

- We were looking for a dataset with relevant information and with a significant size to train the data with ML.

### WHERE?

University of Wisconsin (DS) repository

### FORMAT

CSV file

### SIZE

569 datapoints

### CONTENT

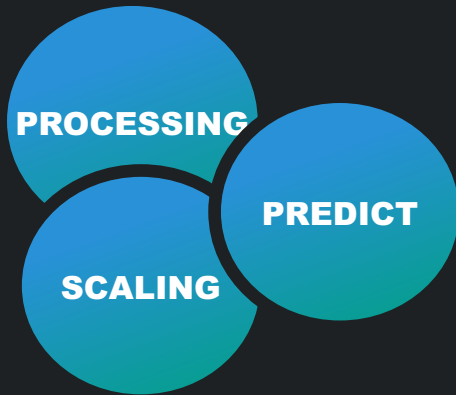
30 variables of characteristics of the tumor and the final diagnosis

## CLEANSING

- 1) Cleansing for database creation
  - Drop of columns without data
  - Check for missing values
- 2) Cleansing for ML model
  - Pre-process of categorical data
  - Drop of unnecessary columns



# ML MODEL DESING



## ● LOGISTIC REGRESSION

	Precision	Recall	F1-score
BENIGN	0.97	0.97	0.97
MALIGNANT	0.96	0.94	0.95
Accuracy	0.97		

## ● RANDOM FOREST

	Precision	Recall	F1-score
BENIGN	0.99	0.98	0.98
MALIGNANT	0.96	0.98	0.97
Accuracy	0.98		

The model is **accurate** for both benign & malignant tumor. Its predictions are nearly always correct with high **precision** scores and the model correctly finds nearly all the true 'malignant tumors' as the **recall** scores were extremely high.

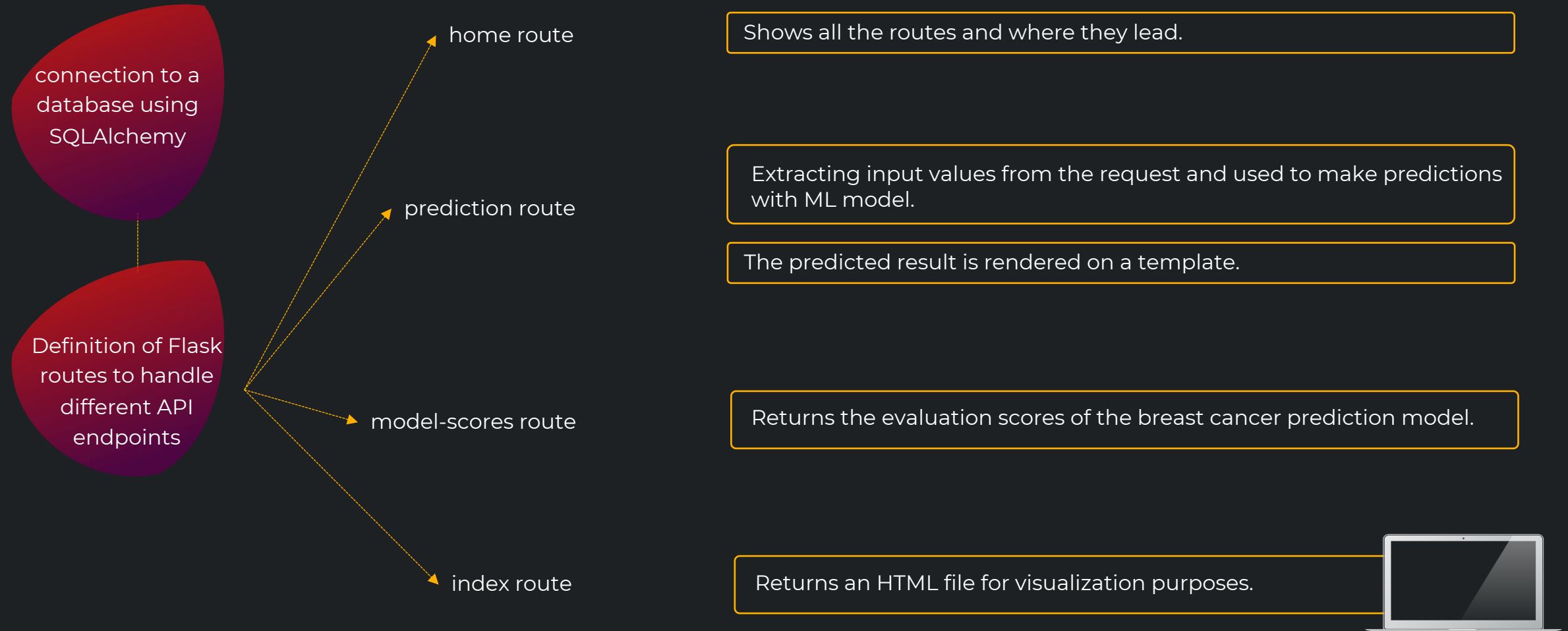
## WHY IT IS IMPORTANT TO LOOK FOR THE HIGHEST PREDICTION RECALL?

The costs of mis-classifying a 'malignant tumor' as a 'benign tumor' are extremely high. It was not acceptable that this kind of tool could misdiagnose a patient with cancer as a 'healthy patient' and send them home without treatment.



# API INTEGRATION

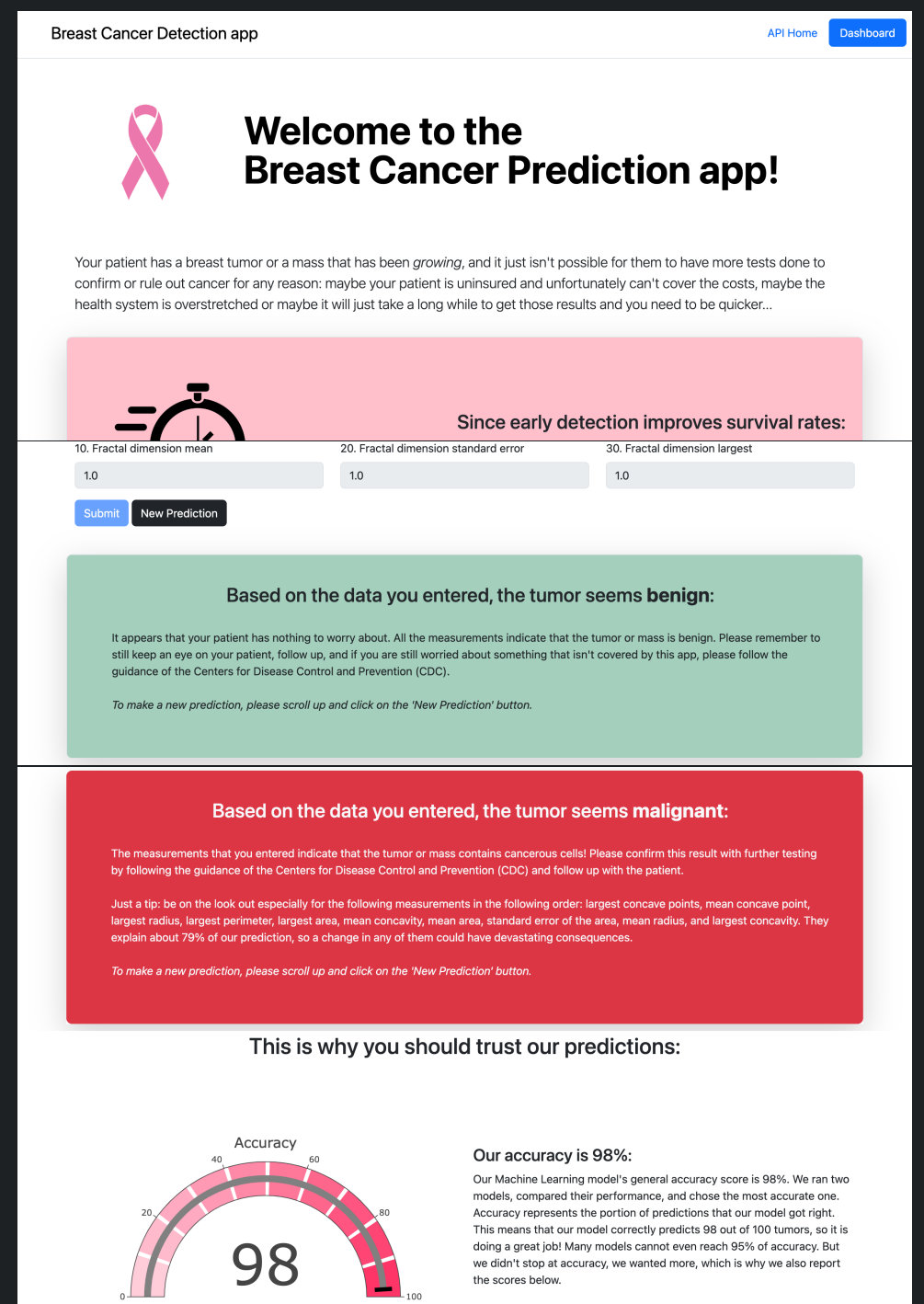
**WHAT?** Flask API that allows users to submit input data for a breast cancer prediction and returns the scores of the ML model.





# VISUALIZATION / DASHBOARD

- **Friendly** and **simple**, empowered by a potent ML model.
- Allows users (doctors) to **make decisions** before recommending more expensive tests
- 30 inputs to fill out and compare the patient's tumor vs. prediction model
- Charts show how trustworthy our model is
- Prediction result and recommendation





## CONCLUSIONS / NEXT STEPS

- Our model was **successful** on learning with the provided dataset and developing a high level of prediction of breast cancer.
- The **random forest** model achieved the highest accuracy, precision, f1-scores and recall vs. logistic regression.
- ML has the potential to **reduce costs** while still being reliable for screening breast cancer.

## NEXT STEPS

- Having **more recent data** could improve our precision and make our model be more in touch with today.
- Having **more feature variables** could improve the model scores or even work better with other ML models, such as a Neural Network.

ANY QUESTIONS?

