

BREAST CANCER PREDICTION USING LOGISTIC REGRESSION

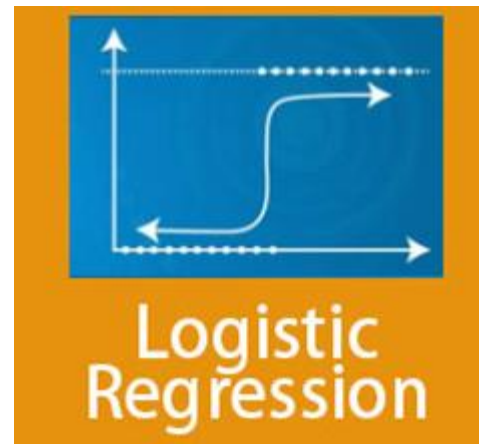
BDA-2 Mini Project

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Overview



1. Dataset: UCI Wisconsin Breast Cancer

2. Database driver: PyMongo

3. Machine Learning: Logistic Regression

4. Big data method: Pyspark

5. Visualization: Tableau





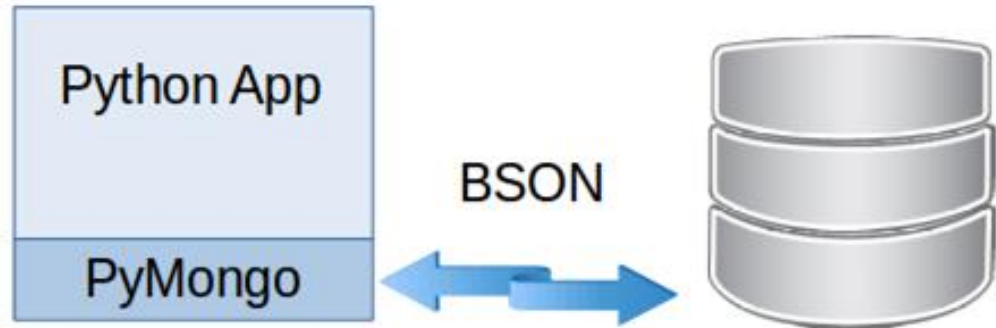
Dataset

- Name: Breast Cancer Wisconsin (Diagnostic) Data Set
- About the dataset:
 - ID number (1)
 - Diagnosis (M = malignant, B = benign) (2)
 - (3-32) Attributes - Ten real-valued features are computed for each cell nucleus:
- Total 30 features.
- Class distribution: 357 benign, 212 malignant

Attributes:

- a) **radius** (mean of distances from center to points on the perimeter)
- b) **texture** (standard deviation of gray-scale values)
- c) **perimeter**
- d) **area**
- e) **smoothness** (local variation in radius lengths)
- f) compactness ($\text{perimeter}^2 / \text{area} \cdot 1.0$)
- g) **concavity** (severity of concave portions of the contour)
- h) **concave points** (number of concave portions of the contour)
- i) **symmetry**
- j) **fractal dimension** ("coastline approximation" - 1)

PyMongo



A MongoDB driver for Python to access the MongoDB database

#10) calculate the total of fractal dimension of patients with Benign cancer and perimeter is greater than 100

```
agr = [{ '$match': {'$or': [ { 'diagnosis': "B" }, { 'perimeter_mean': { "$gt": 100 } } ] }},  
        { '$group': { '_id': 1, 'total': { '$sum': "$fractal_dimension_mean" } } }]  
val = list(db.cancer.aggregate(agr))  
  
print('The requested value is {}'.format(val[0]['total']))
```

The requested value is 31.9960899999999985

Logistic Regression

- Logistic Regression is commonly used to estimate the probability that an instance belongs to a particular class.

- In our case, what is the probability that this tumor is malignant or benign?

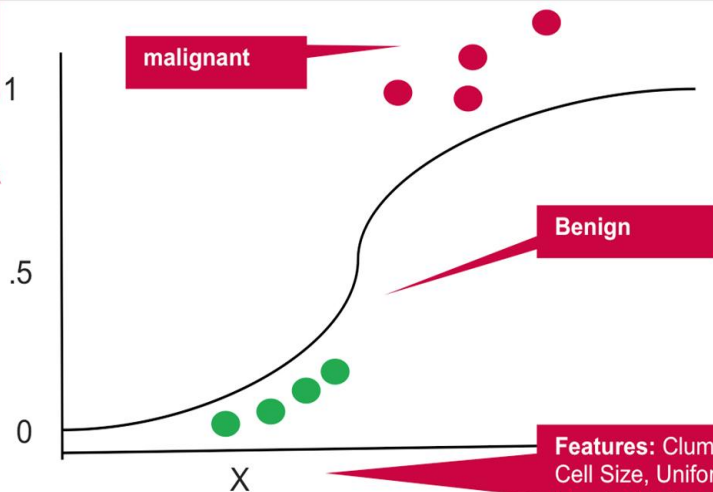
- Binary Classifier

Estimated Probability -

> 50% Positive Class (1 / Malignant)

< 50% Negative Class (0 / Benign)

Label
Probability
Malignant



malignant

Benign

Features: Clump Thickness, Uniformity of Cell Size, Uniformity of Cell Shape, Marginal Adhesion, Single Epithelial Cell Size, Bare Nuclei, Bland Chromatin, Normal Nucleoli, Mitoses

Hadoop vs Spark

Parameters	Hadoop	Spark
Performance	Less optimal	More optimal
Latency	High latency computing	Low latency computing
Data	Process data in batch mode	Can process interactively
Usage	Batch processing with a huge volume of data	Process real-time data, from real-time events like Twitter, Facebook
Scheduler	External job scheduler is required	In memory-computation, no external scheduler required
Ease of use	Model is complex, need to handle low-level API's	Easier to use, abstraction enables a user to process data with high-level operators
Security and Cost	Highly secure and less costly since MapReduce model provides a cheaper strategy	Less secure and costlier since it requires in-memory solution

Why use PySpark?

1. Requirements

- a. Data coming in from multiple systems.
- b. Real-time as well as batch data.
- c. We need to perform data analytics over all these data inputs by building a system combining it.

2. Why not Pandas?

- a. Pandas is great for tabular data with millions of rows.
- b. Many features compared to PySpark.
- c. Limitations - Distributed data and/or real-time data.

3. Note

- a. Our dataset is not distributed or even real-time.
- b. While we have used PySpark module, there is no significant difference in efficiency and execution times, compared to a module without PySpark.

Tableau

PROS

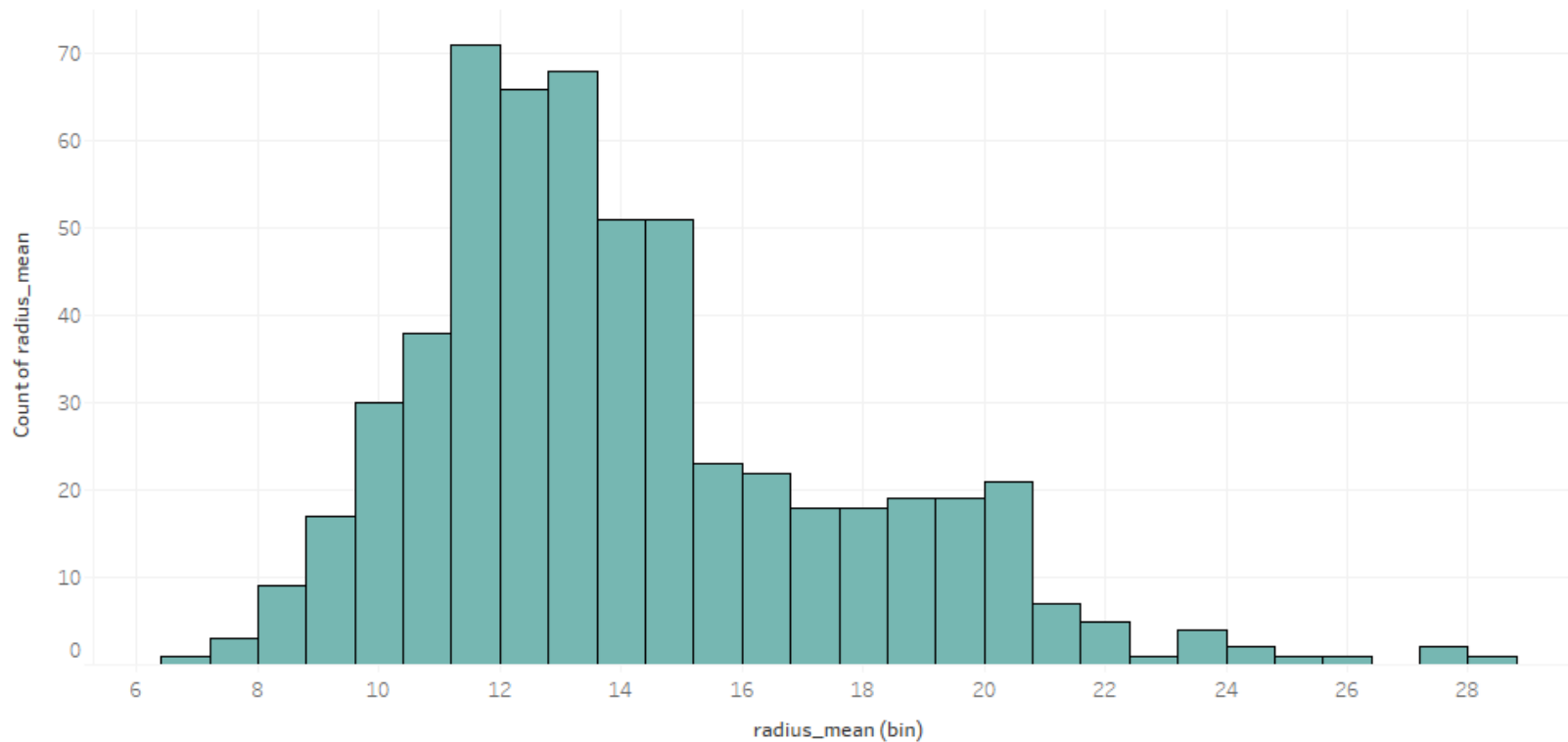
- Super easy to learn to use
- Ability to do complex analysis
- Great for data exploration - when you're not sure what you're looking to build
- Relatively short development time - anywhere from 1 hour to a few days depending on the complexity
- Fast data engine - does not require a database hit every time
- Great online community/support
- Tableau Server/Online provide easy sharing platform

CONS

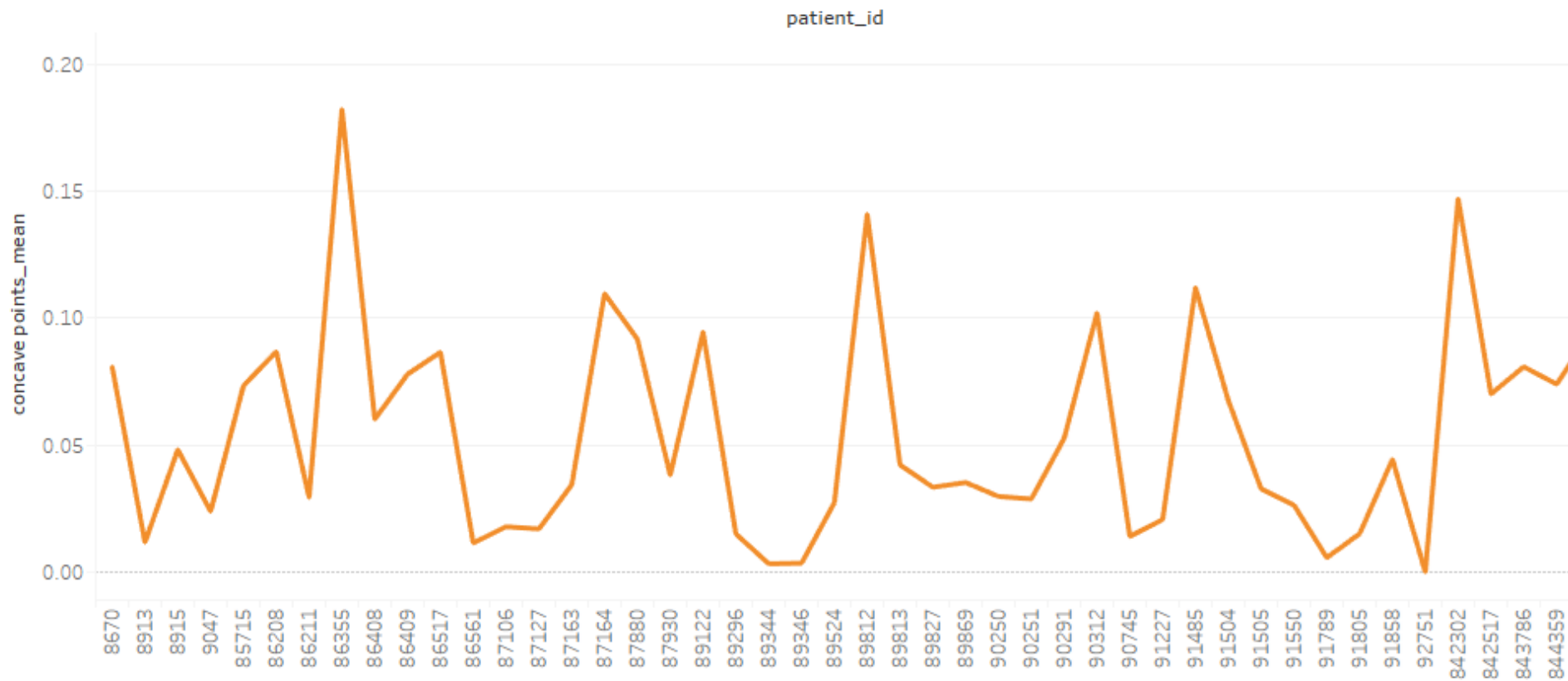
- Can be expensive for mass-consumption (unless you're using Tableau Public)
- Not easy to integrate app-development on top of (though this is getting better with the APIs)

Visualization

Radius Histogram

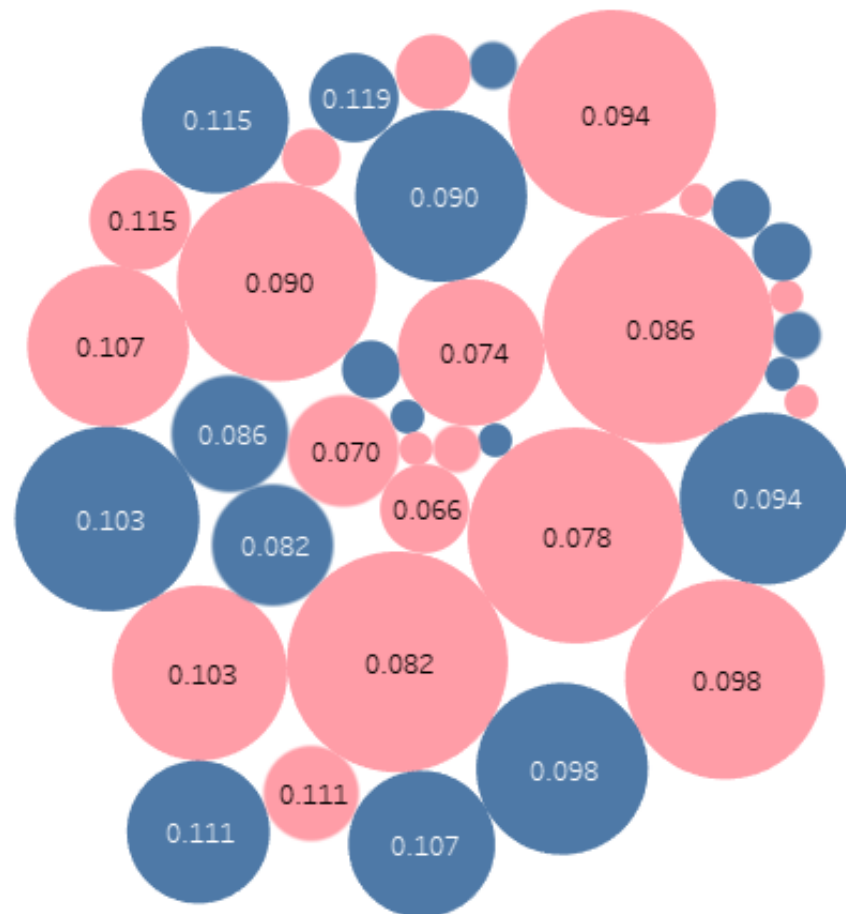


Line Graph showing concave points of each patient:



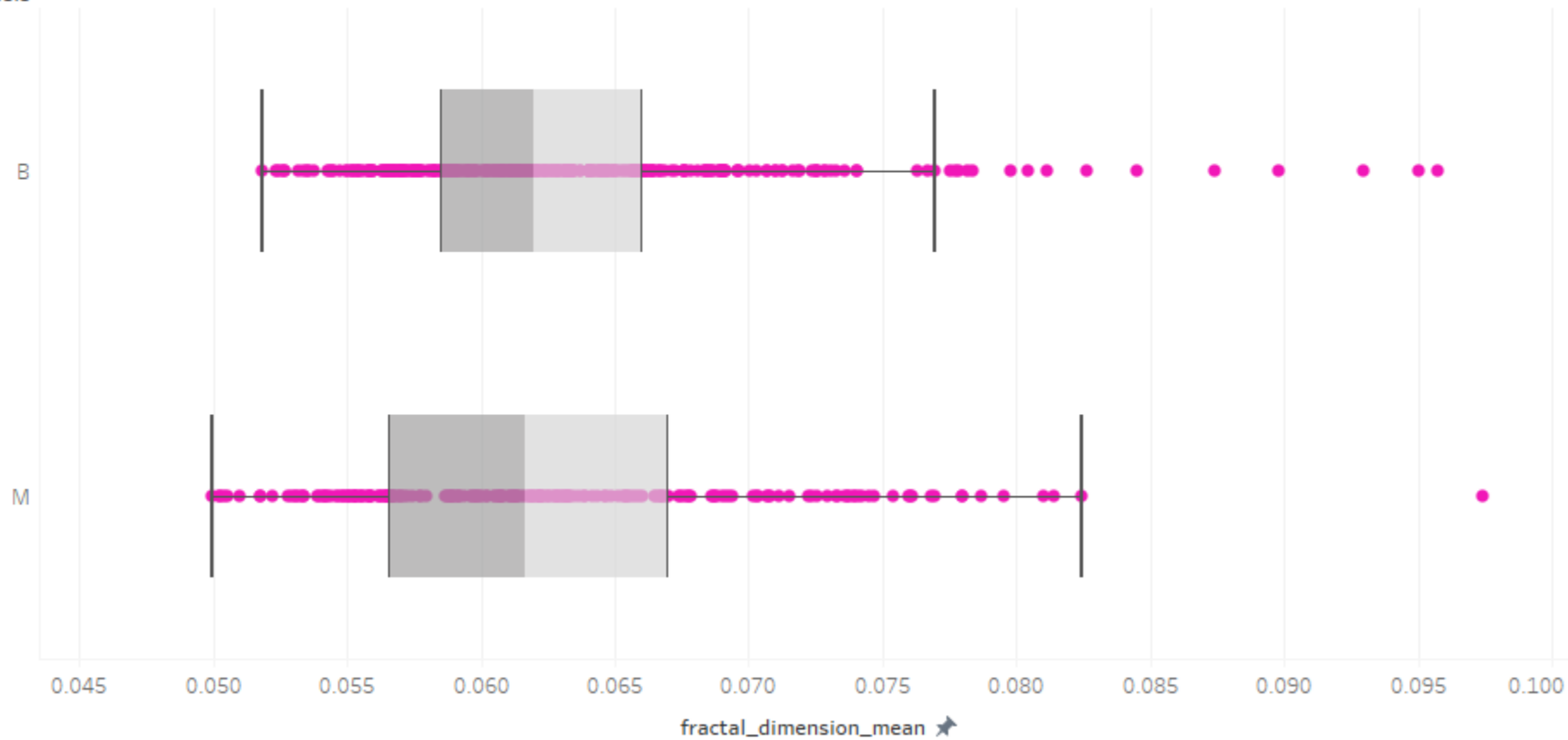
Bubble Chart

(colour:- **Malignant**, **Benign**; size:-no. of people diagnosed, label:- smoothness_index)

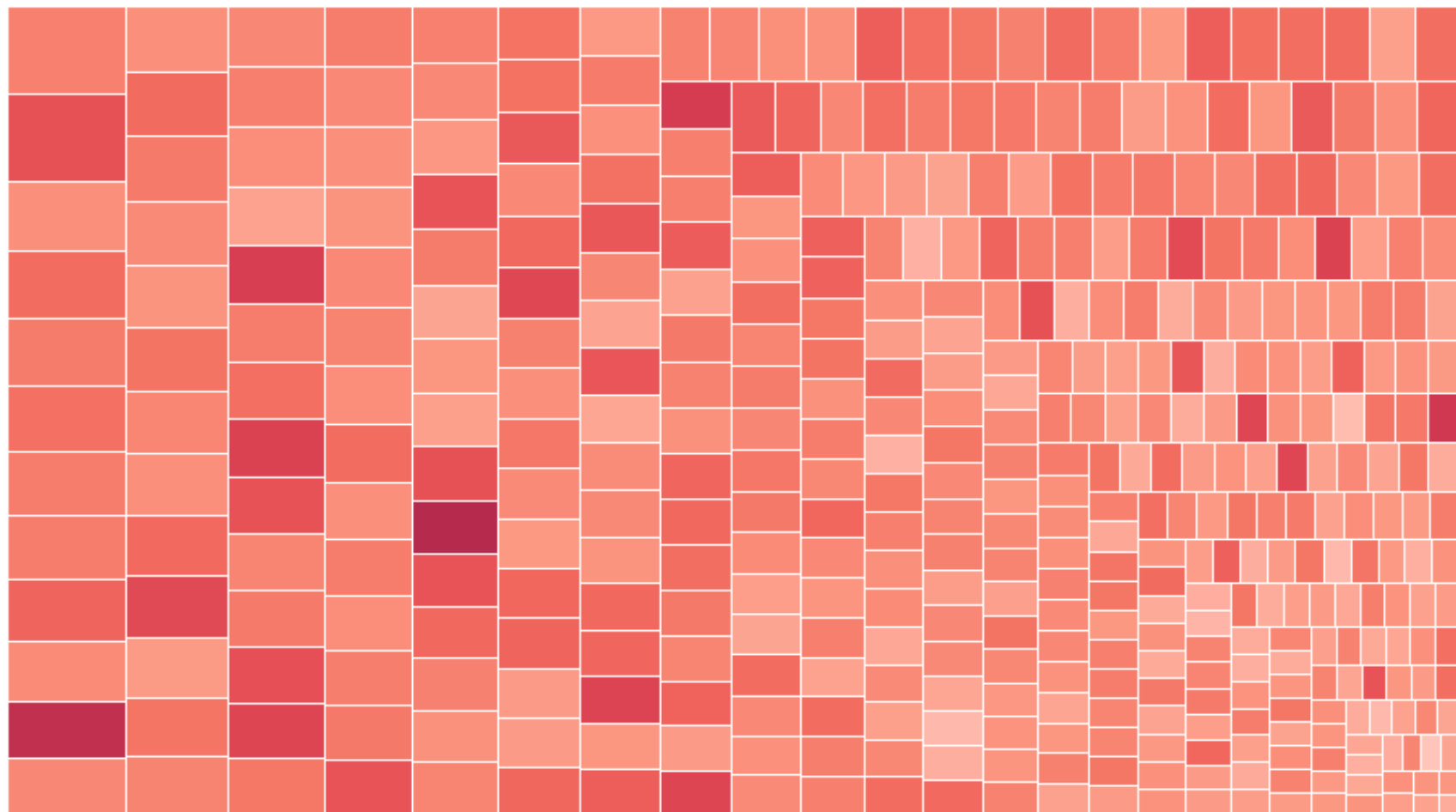


Box Plot for outlier detection of the *fractal-dimension*:

diagnosis



Tree Map (darker colour=more symmetry, greater size=more compactness)

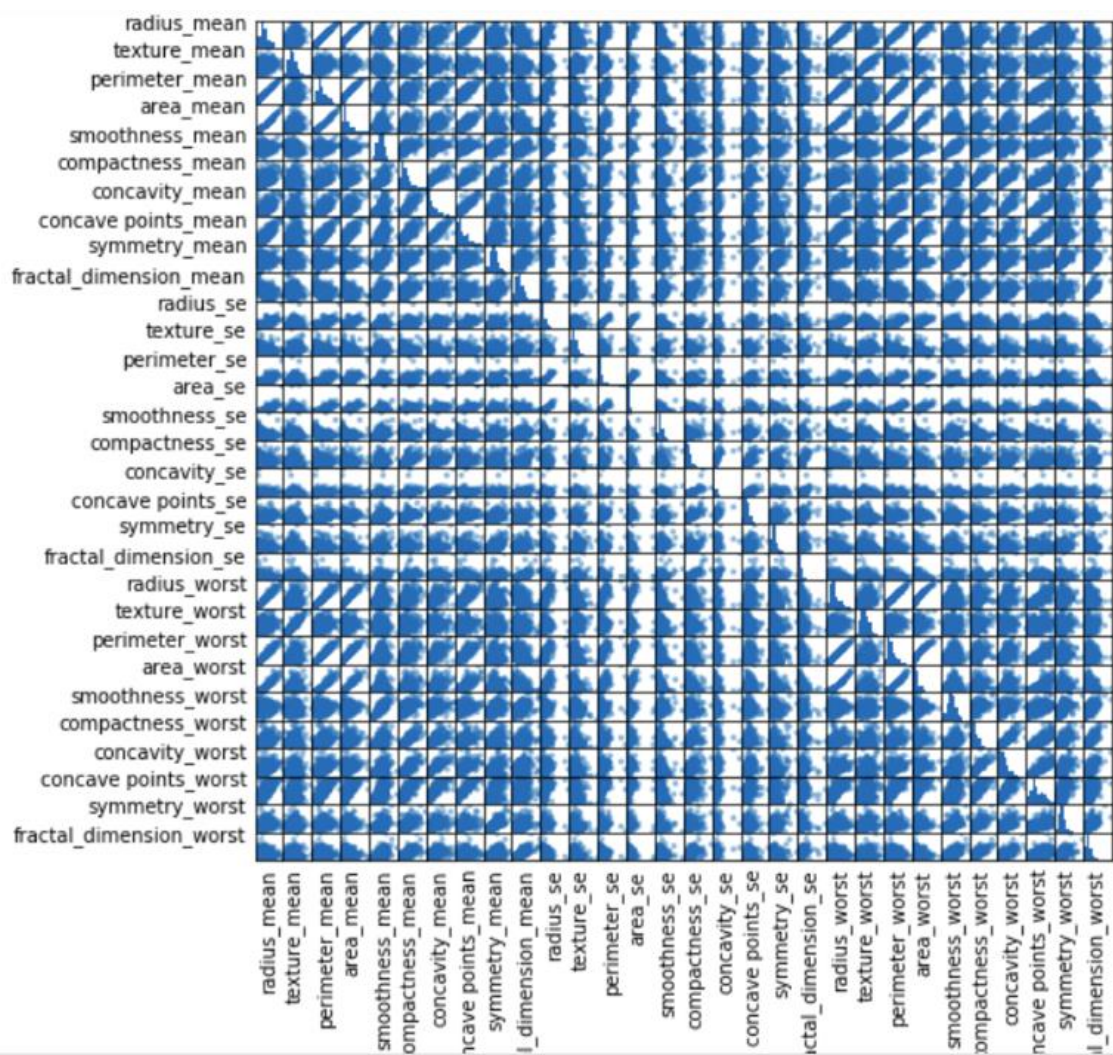


Scatter Plot : Area vs Texture (Benign⁺, Malignant⁺)



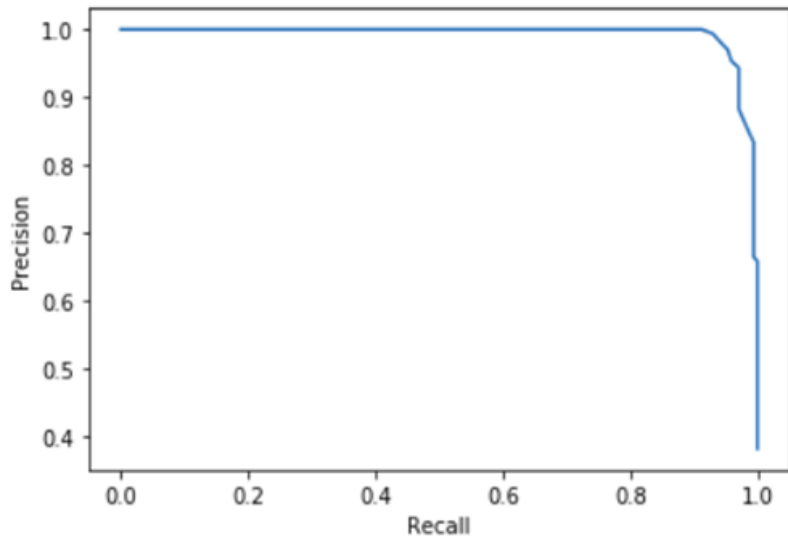
Scatter Matrix

— — —

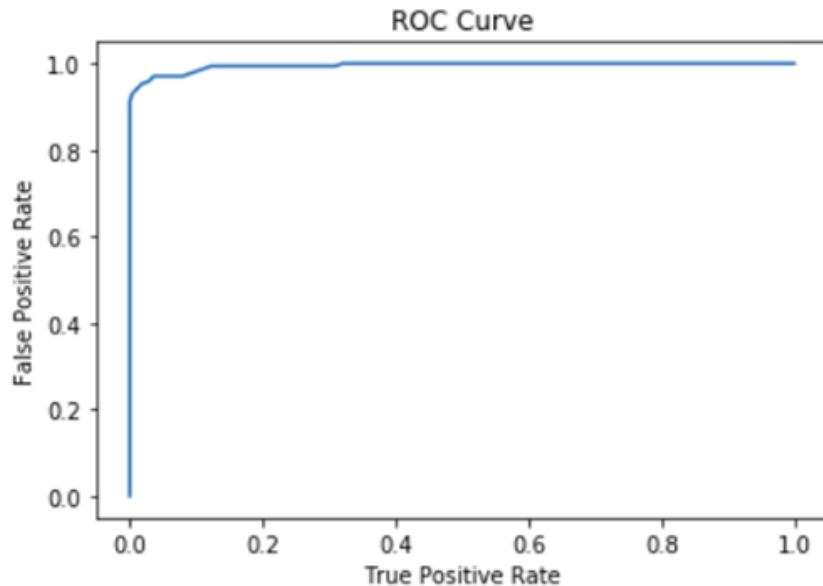


Performance

Precision vs Recall



ROC curve for training data



Training set areaUnderROC: 0.9950039968025579

Results

```
1 predict_test=model.transform(test)
2 predict_test.select("label","prediction").show()
```

label	prediction
1.0	1.0
1.0	1.0
0.0	0.0
1.0	1.0
1.0	1.0
0.0	0.0
0.0	0.0
0.0	0.0
1.0	1.0
1.0	1.0
1.0	1.0
1.0	1.0
0.0	0.0
0.0	0.0
0.0	0.0
1.0	1.0
1.0	1.0
0.0	0.0
0.0	0.0
0.0	0.0

only showing top 20 rows

← **Label & Predictions**

Area under ROC for Test set

↓

```
from pyspark.ml.evaluation import BinaryClassificationEvaluator
evaluator=BinaryClassificationEvaluator(rawPredictionCol='rawPrediction',labelCol='label')
predict_test.select("label","rawPrediction","prediction","probability").show(5)
print("The area under ROC for train set is {}".format(evaluator.evaluate(predict_train)))
print("The area under ROC for test set is {}".format(evaluator.evaluate(predict_test)))
```

label	rawPrediction	prediction	probability
1.0	[-9.1119950415597...	1.0	[1.10322201416982...
1.0	[-1.3667099321543...	1.0	[0.20315192649587...
1.0	[-5.7255420138824...	1.0	[0.00325098161966...
1.0	[-7.8215833886522...	1.0	[4.00825537245302...
1.0	[-5.5248774023552...	1.0	[0.00397052943089...

only showing top 5 rows

The area under ROC for train set is 0.9952010871873272

The area under ROC for test set is 0.997925925925926



Let's defeat
**breast
cancer**
together