

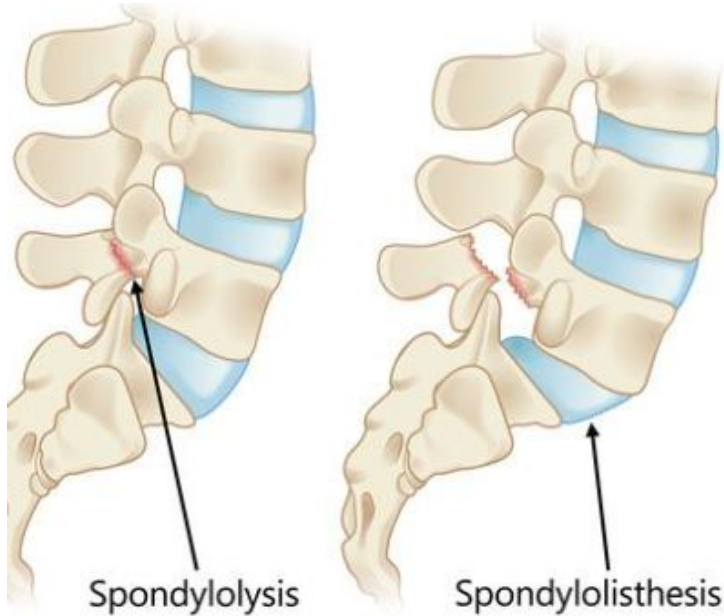
Springboard Capstone 1

~~~Data Science Career Track~~~  
Biomechanical Features of Orthopedic Patients

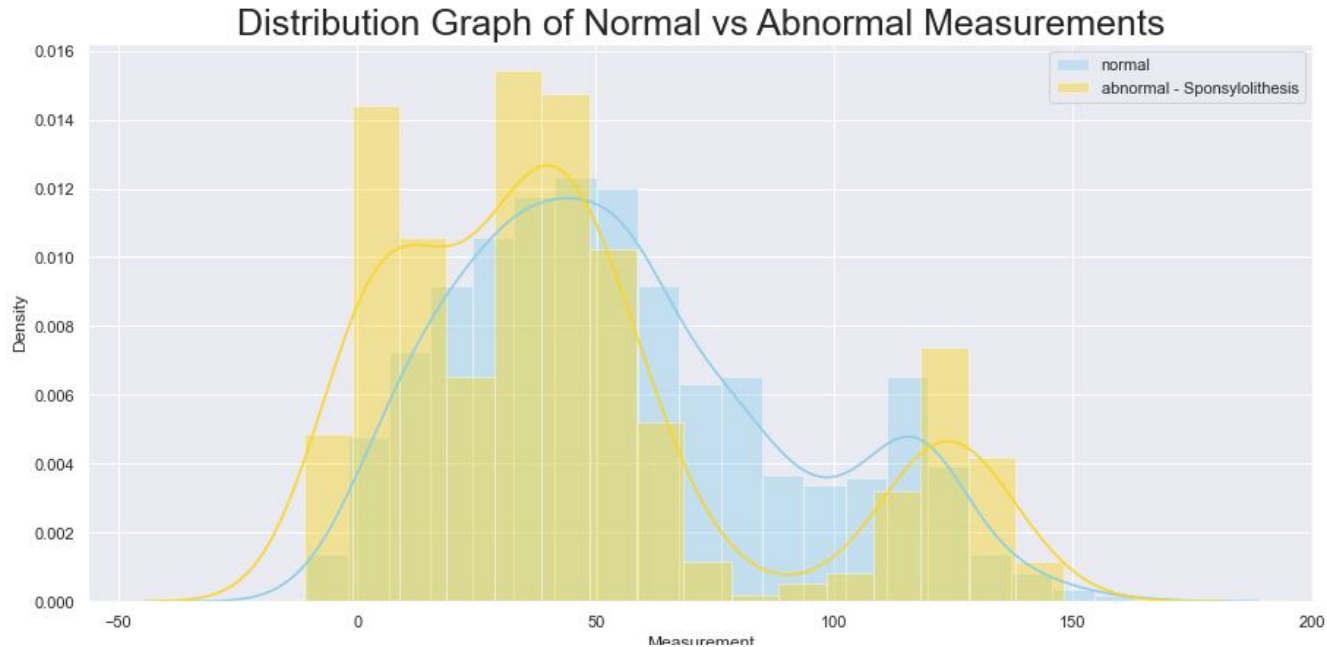
CLASSIFICATION  
of  
Spondylolisthesis

# Painful lumbar (back) issues are primarily diagnosed on x-ray results

The problem is - Not all issues are as obvious as seen below

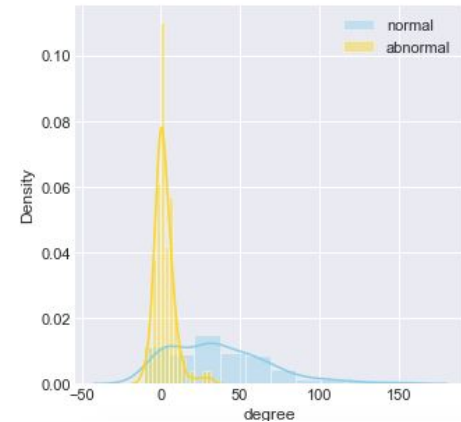
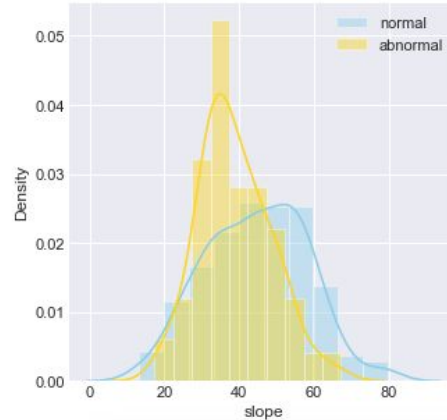
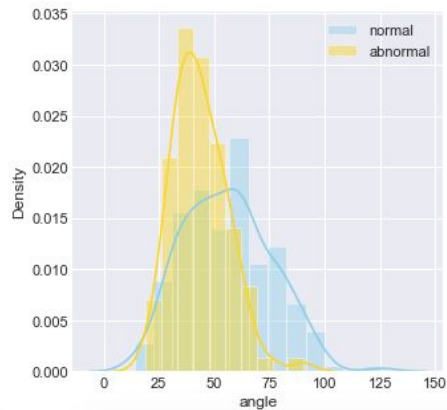
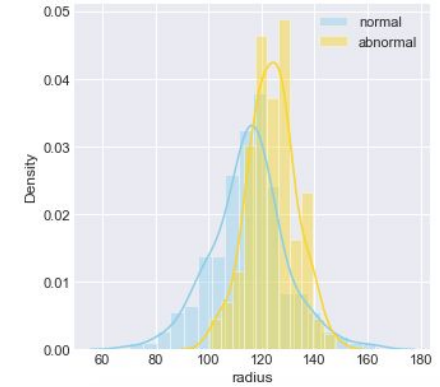
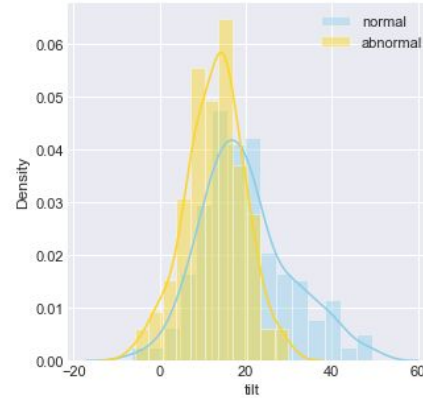
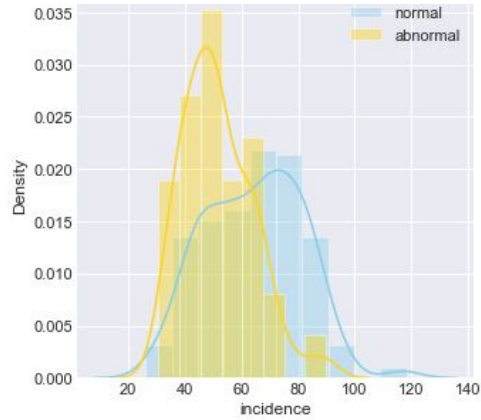


*Abnormal measurements overlap Normal measurements.*



It appears the high-bills you've paid for radiology results may have been well earned.

Proper diagnosis requires comparison of several features - *simultaneously*



The physician bears the stress to simultaneously compare multiple measurements.

| pelvic_incidence | pelvic_tilt numeric | lumbar_lordosis_angle | sacral_slope | pelvic_radius | degree_spondylolisthesis | class    |
|------------------|---------------------|-----------------------|--------------|---------------|--------------------------|----------|
| 63.027818        | 22.552586           | 39.609117             | 40.475232    | 98.672917     | -0.254400                | Abnormal |
| 39.056951        | 10.060991           | 25.015378             | 28.995960    | 114.405425    | 4.564259                 | Abnormal |
| 68.832021        | 22.218482           | 50.092194             | 46.613539    | 105.985135    | -3.530317                | Abnormal |
| 69.297008        | 24.652878           | 44.311238             | 44.644130    | 101.868495    | 11.211523                | Abnormal |
| 49.712859        | 9.652075            | 28.317406             | 40.060784    | 108.168725    | 7.918501                 | Abnormal |

Machines are perfectly suited to assist.

# Project: Classification of Spondylolisthesis

## Problem:

Diagnosis requires simultaneous comparison of multiple features.

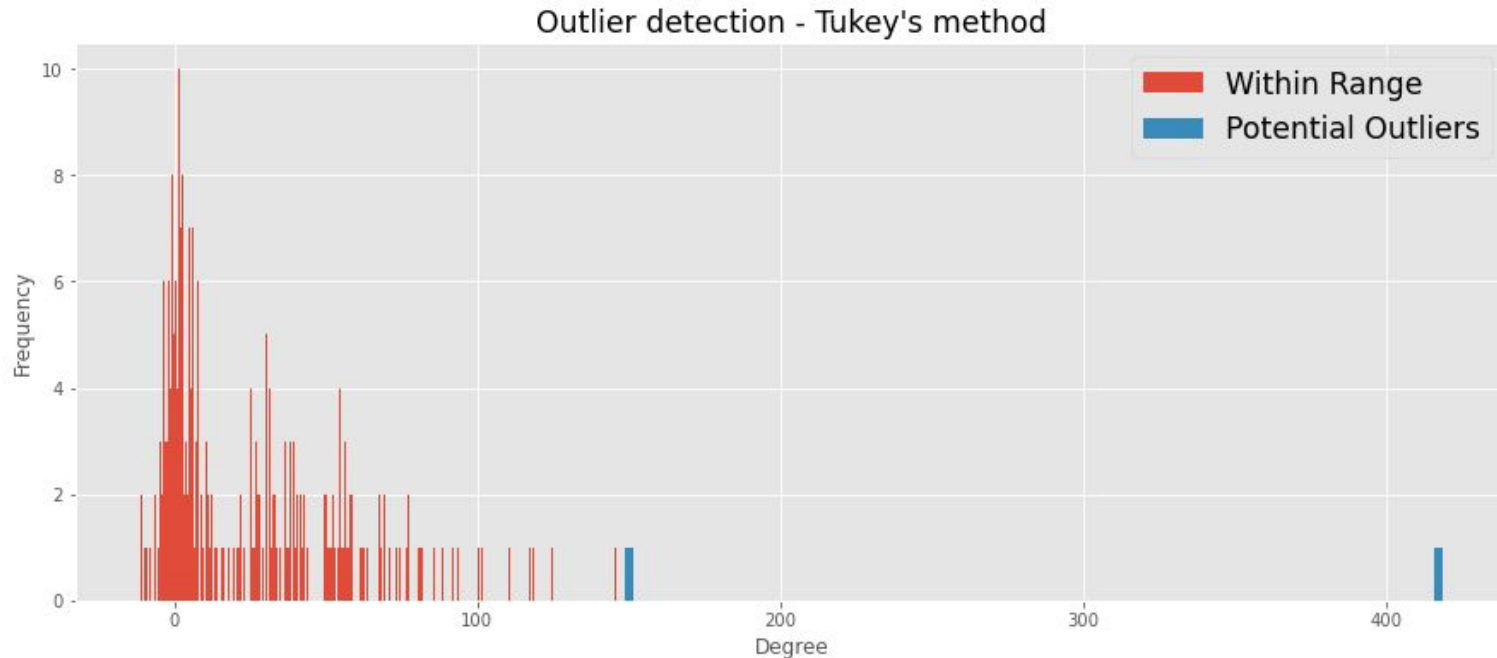
## Solution:

Machine learning quickly classifies and sorts saving time and improving accuracy.

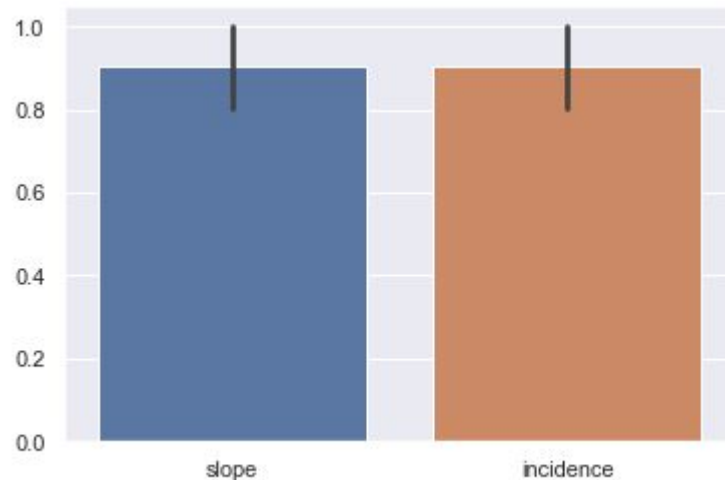
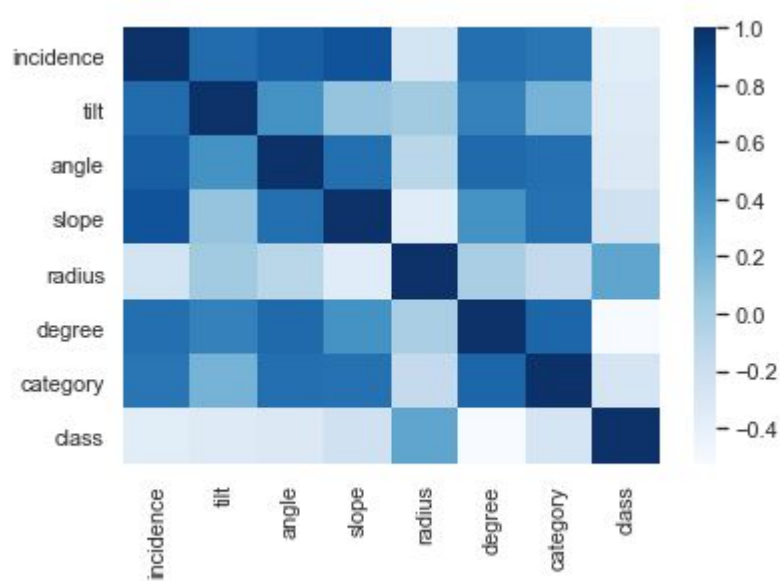
# The Data

'Tukey's' algorithm was used to locate ***potential*** outliers.

1 outlier (410) was >2 times as large as the next data point (150) - it was removed.



## Correlations plotted to look for overly-correlated features



**'slope' and 'incidence' were highly correlated, close to the 90% cut-off range.**

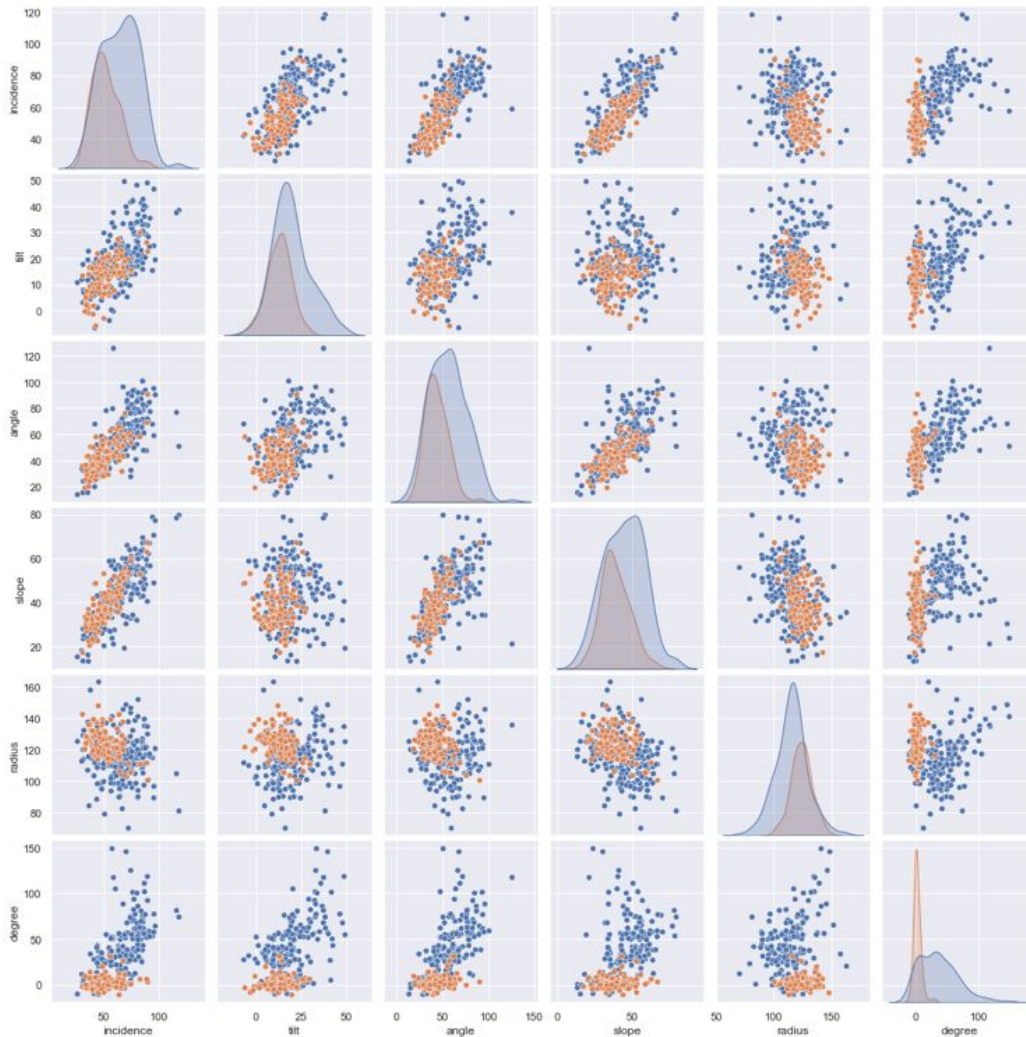


**A correlation module ran to remove overly-correlated features (>90%). All passed, none removed.**

# Correlations of features

Comparing the separation between "Normal" (orange) and "Abnormal" (blue).....

suggests that added weight might be useful on 'degree' and 'radius' to improve the model.



# The Models

4 Models were used

- Logistic Regression
- Gaussian Naive Bayes
- KNearest Neighbors
- Random Forest

# Logistic Regression & The Base

A basic Logistic Regression was used to determine the base accuracy of 0.87

Additional tuning of C parameter using GridSearch did not change the results

# Gaussian Naive Bayes

Due to the bi-modal and right-skewed normal shape of some features - a Gaussian NB was tested with results less than the base

# KN Neighbors

KN Neighbors returned the same accuracy after tuning the `n_neighbors` to =10

Tuning of parameters: `algorithm`, `leaf_size`, `metric`, `metric_params` did not improve accuracy.

# Random Forest

Random Forest provided improved results of 88% once the `n_estimator` and `random_state` were tuned in tandem.

A module was created to loop through a range of values, by plotting the results we see the optimal values.

