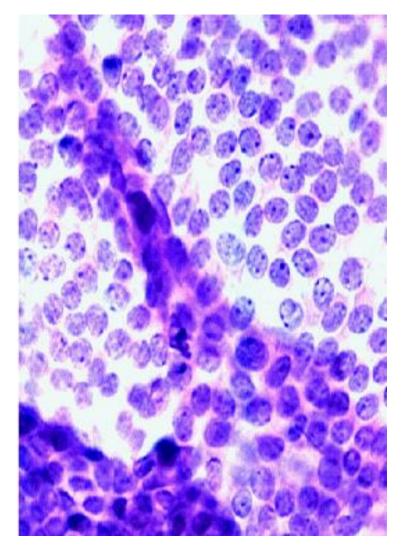
# XGBClassifier Most Favorable for Breast Cancer Diagnosis

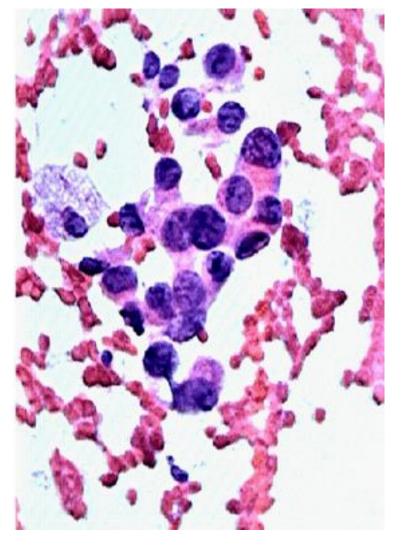
Rachel Khoo

Thinkful Capstone 2

September 2020

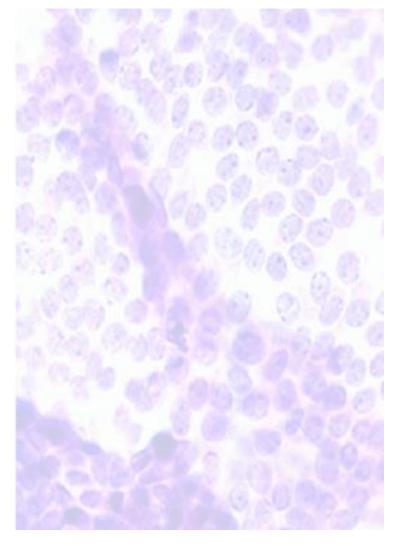
## Who has cancer?

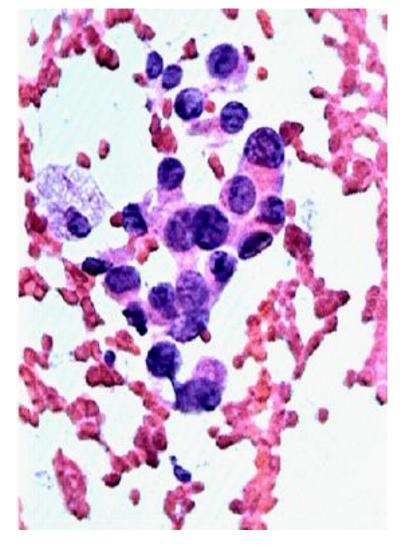




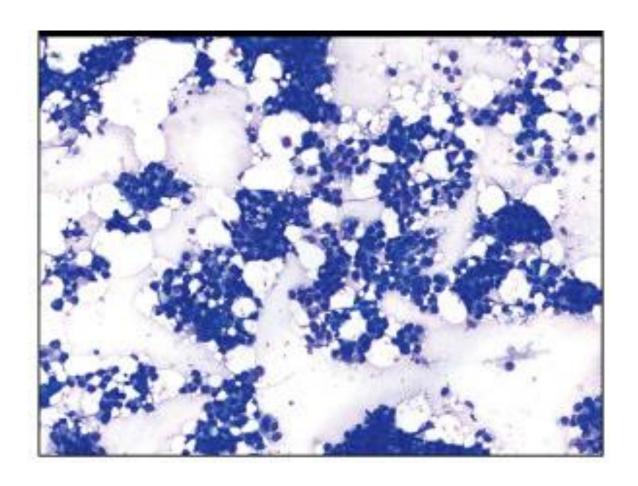
Eickhoff, Carsten. (2014). Crowd-powered experts: helping surgeons interpret breast cancer images. ACM International Conference Proceeding Series. 53-56. 10.1145/2594776.2594788.

## Who has cancer?

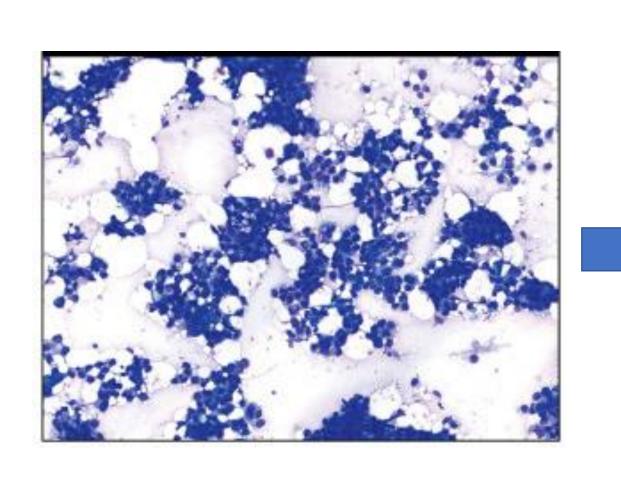




## Who has cancer?



# Machine Learning can make diagnosis easier



• Area: 1001.0

• Texture: 10.38

• Compactness: 0.27760

• Concavity: 0.3001

## Machine Learning can make diagnosis easier

• Area: 1001.0

• Texture: 10.38

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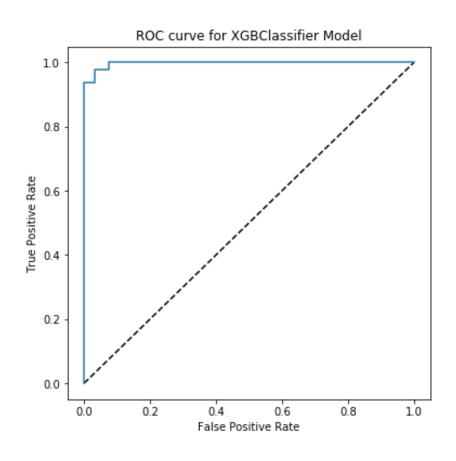
#### XGBClassifier is the best model

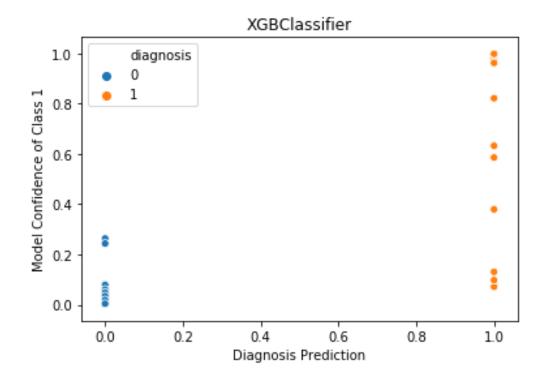
• AUPRC: 0.9699

• Recall: 90%

	Predicted Benign	Predicted Malignant
True Benign	68	4
True Malignant	4	38

## XGBClassifier is confident and accurate





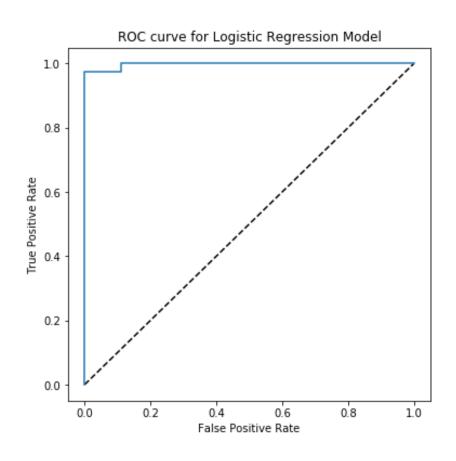
# Logistic Regression: Accuracy isn't everything

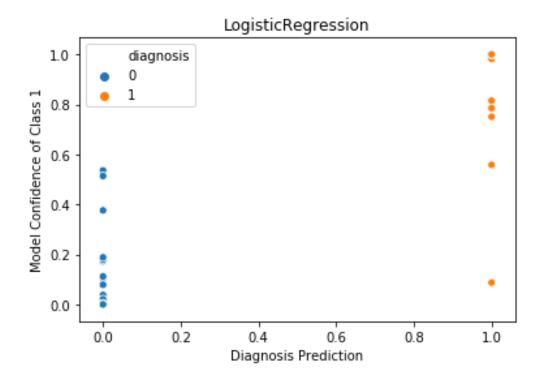
• AUPRC: 0.9466

• Recall: 93%

	Predicted Benign	Predicted Malignant
True Benign	71	1
True Malignant	3	39

# Logistic Regression is more confidently wrong





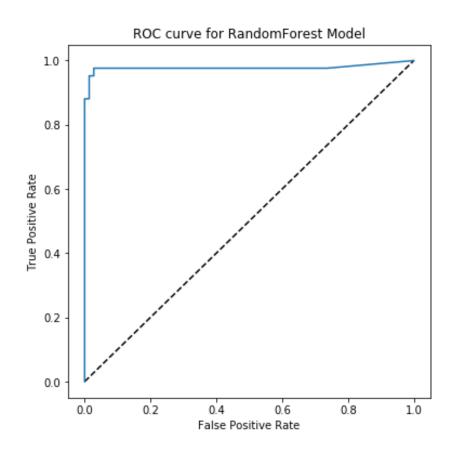
# Random Forest: good accuracy, bad recall

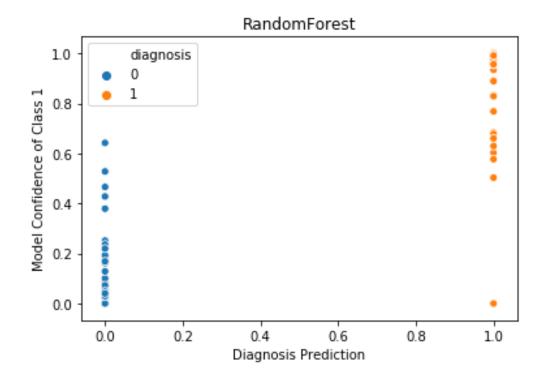
• AUPRC: 0.9395

• Recall: 88%

	Predicted Benign	Predicted Malignant
True Benign	71	1
True Malignant	5	37

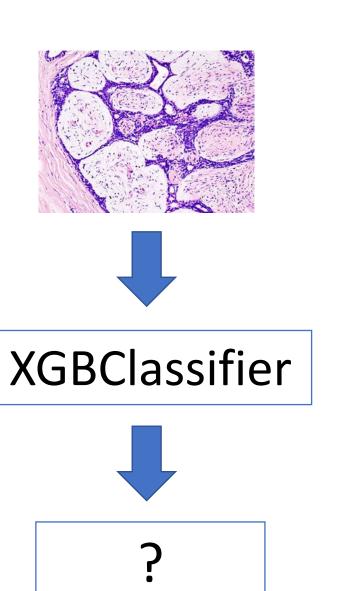
## RandomForest can still be useful





# Every model has limitations

- XGB
  - Can't predict outside of sample
- Logistic Regression
  - Can be slow
- KNN
  - Slower
  - Not easily interpretable
  - Can't predict outside of sample

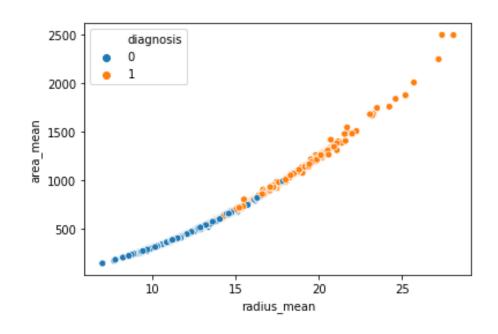


# How can we improve accuracy?

More data

Spend more time tuning hyperparameters

PCA to reduce complexity redundancy



#### An ensemble method could be even better

