## A Data Scientist's Value: Are you worth more than the median Salary?

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#### Who should care?

- DS Employers trying to evaluate their workers
- DS Job-seekers trying to figure out how best to stand out
- DS Students wondering which major will help most with a Data Science career

## Problem: Can these predict salary?

- 1. Gender (i.e., Male or Female)
- 2. Age
- 3. Education (i.e., Bachelors, Masters, or PhD)
- 4. Experience (i.e., # of years in current position)
- 5. Major
- 6. Industry

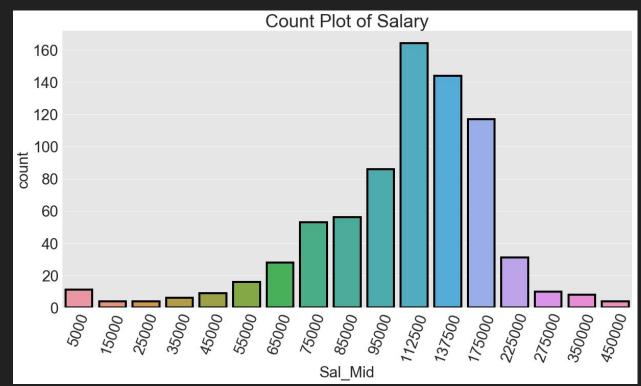
## Data: 2018 Kaggle ML & DS Survey

Original Dataset: 23,859 rows and 395 columns
This Capstone: 751 rows and 7 columns

#### Why? Limited to:

- (1) Residents of the USA,
- (2) Share their salary details, and
- (3) Are employed as Data Scientists

# Salary Distribution for Kaggle Data Scientists



## Hypothesis'

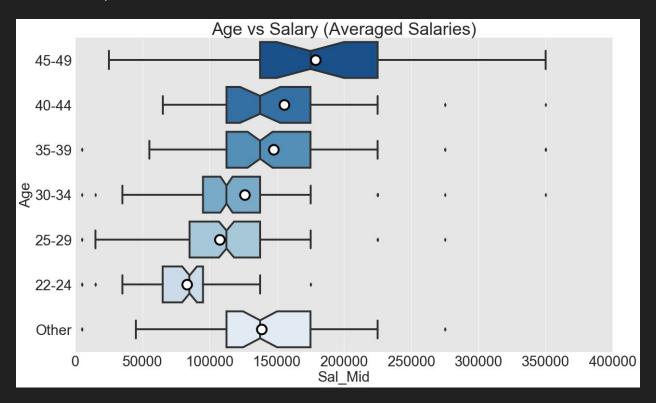
#### Positively Effect

- 1) Older Age
- 2) More Experience
- 3) STEM Majors
- 4) PhD

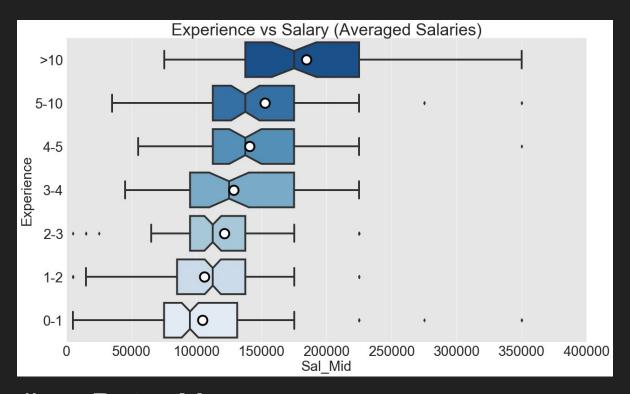
#### **Negatively Effect**

- 1) Younger Age
- 2) Less Experience
- 3) non-STEM Majors
- 4) Bachelor's

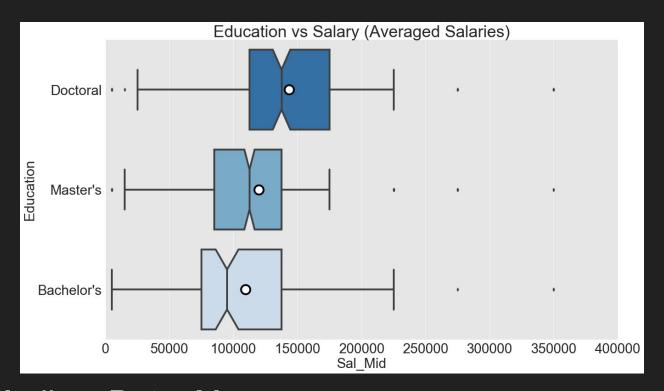
#### The Older, The Richer



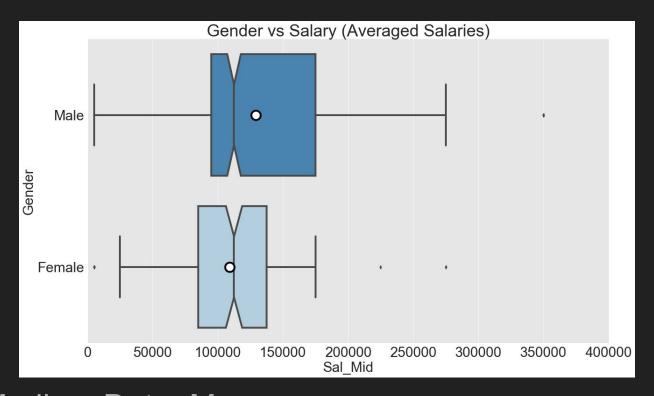
## More Experience, More Salary



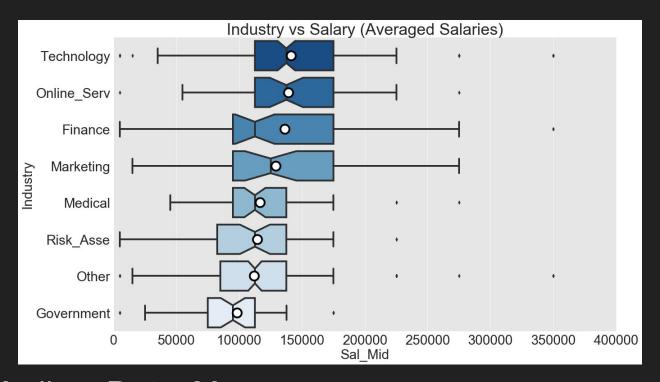
## PhD = Higher Salary



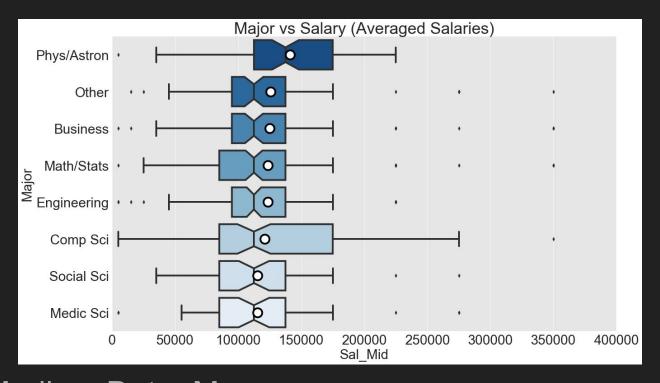
## Male? Higher Salary



## A Trumped Government Does Worst



## Major did not correlate with Salary



#### Results

#### Positively Effect

- Older Age
- 2) More Experience
- 3) STEM Majors
- 4) PhD
- 5) Male\*

#### **Negatively Effect**

- 1) Younger Age
- 2) Less Experience
- 3) non-STEM Majors
- 4) Bachelor's
- 5) Female\*

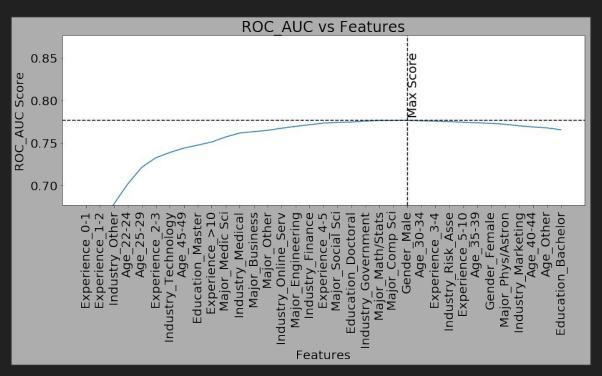
<sup>\*</sup> Simpson's paradox did not appear to be at play.

## Machine Learning

| Classifier               | ROC AUC Score | Best Parameters  |
|--------------------------|---------------|--|
| Logistic Regression      | 0.729         | C = 0.1  |
| K-Nearest Neighbor       | 0.677         | n_neighbors = 12   |
| Random Forest            | 0.708         | criterion = 'entropy', max_depth = 3, max_features = 'auto', n_estimators = 30 |
| Gaussian Naive Bayes     | 0.734         | var_smoothing = 0.1  |
| Xtreme Gradient Boosting | 0.704         | alpha = 10, lambda = 10, max_depth = 2   |

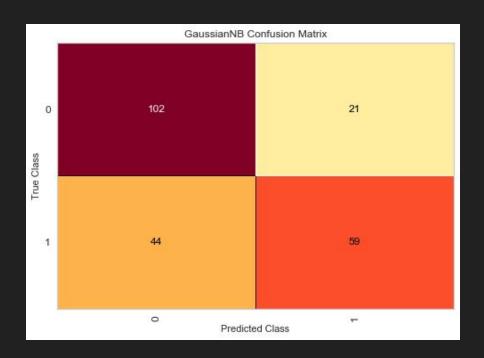
#### Best model was Gaussian Naive Bayes

## Forward Selection Stepwise Regression



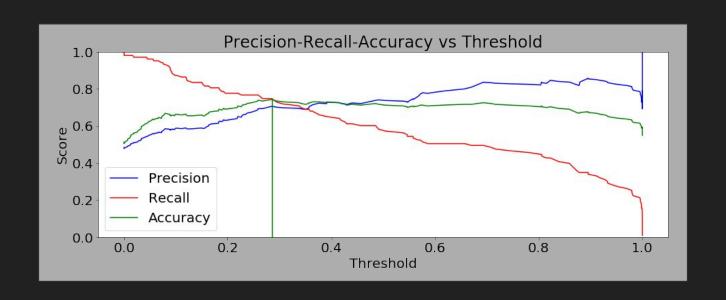
Forward selection used for feature reduction

#### **Confusion Matrix**



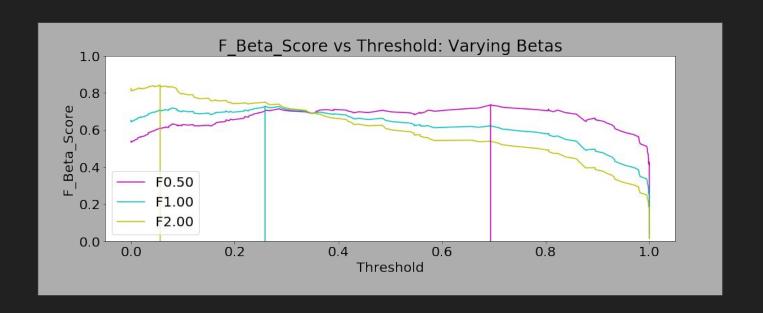
Large number of false negatives ~= true positives

#### P, R & ACC, vs Threshold



Benchmarks used for measuring model usefulness

#### F-Beta Scores vs Threshold



Varying Beta gives more weight to precision or recall

## Business Case Example

Background: Economic hardship

Problem: Not feasible to pay employee desired salary or give raises, but it is also dangerous to pay them below their market value as they may leave.

Solution: More weight to precision; F 0.5 achieves its maximum score of 0.73 when the threshold is set to 0.69.

#### Actionable Recommendations

- DS Employers: Value women equally; Stop asking for Quantitative Majors.
- DS Job-seekers: Government typically pays lower than other sectors; avoid if a high salary is desired.
- DS Students: If in doubt, go with Physics.