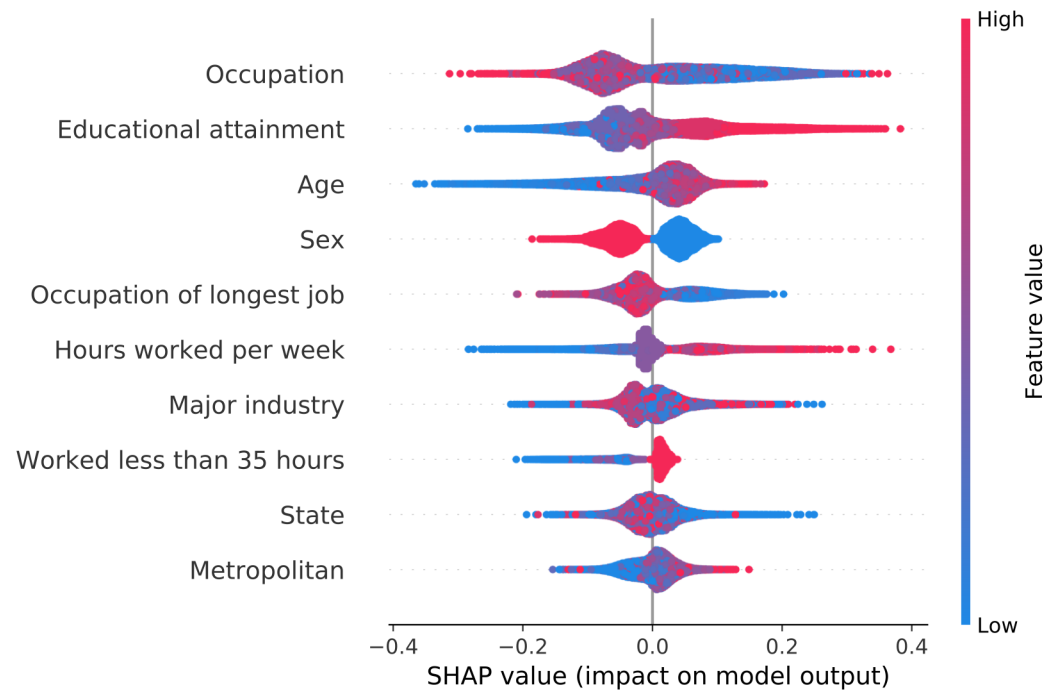
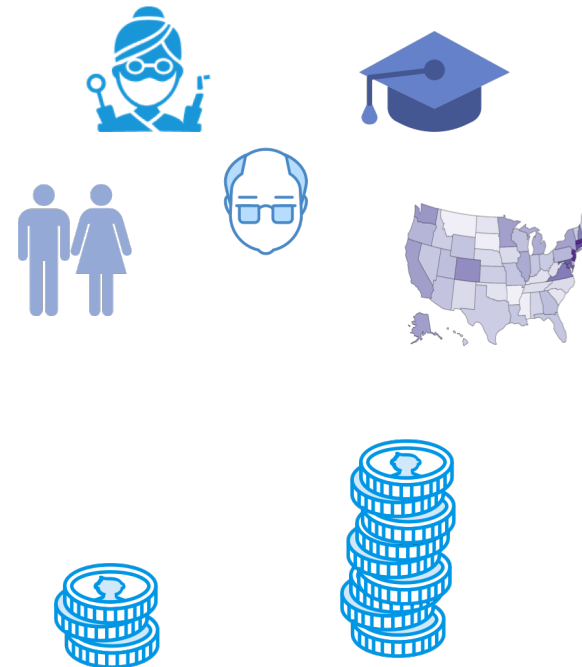


What affects our salary. Analysis and predictions based on the 2007 CPS



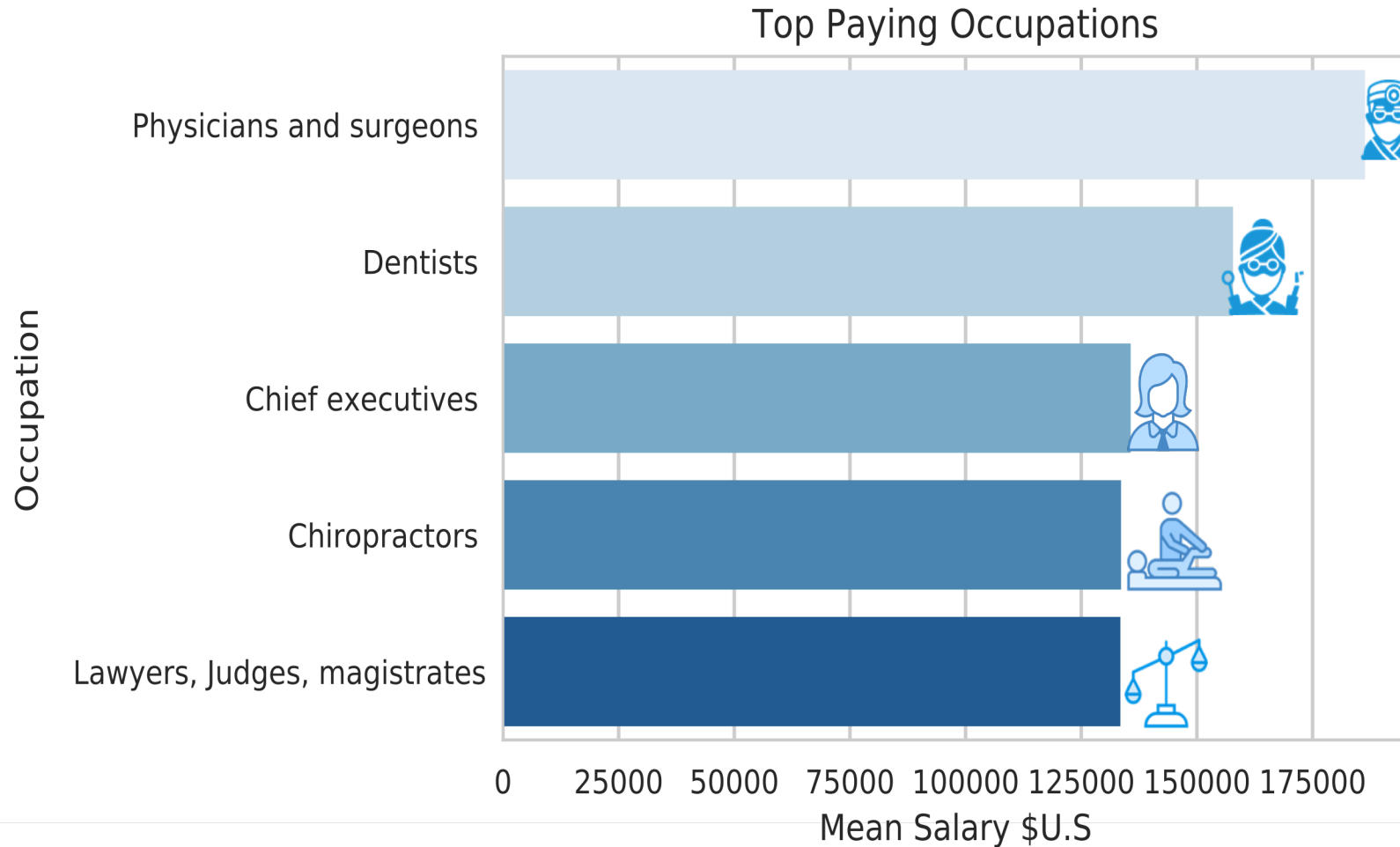
Project Goals

- ▶ Statistical analysis of the 2007 *United States Department of Commerce Current Population Survey*
- ▶ Identifying importance features which impact salary (for example: Occupation, age, gender etc.)
- ▶ Creating a classifier which to identify individuals which earns less then \$40K per year



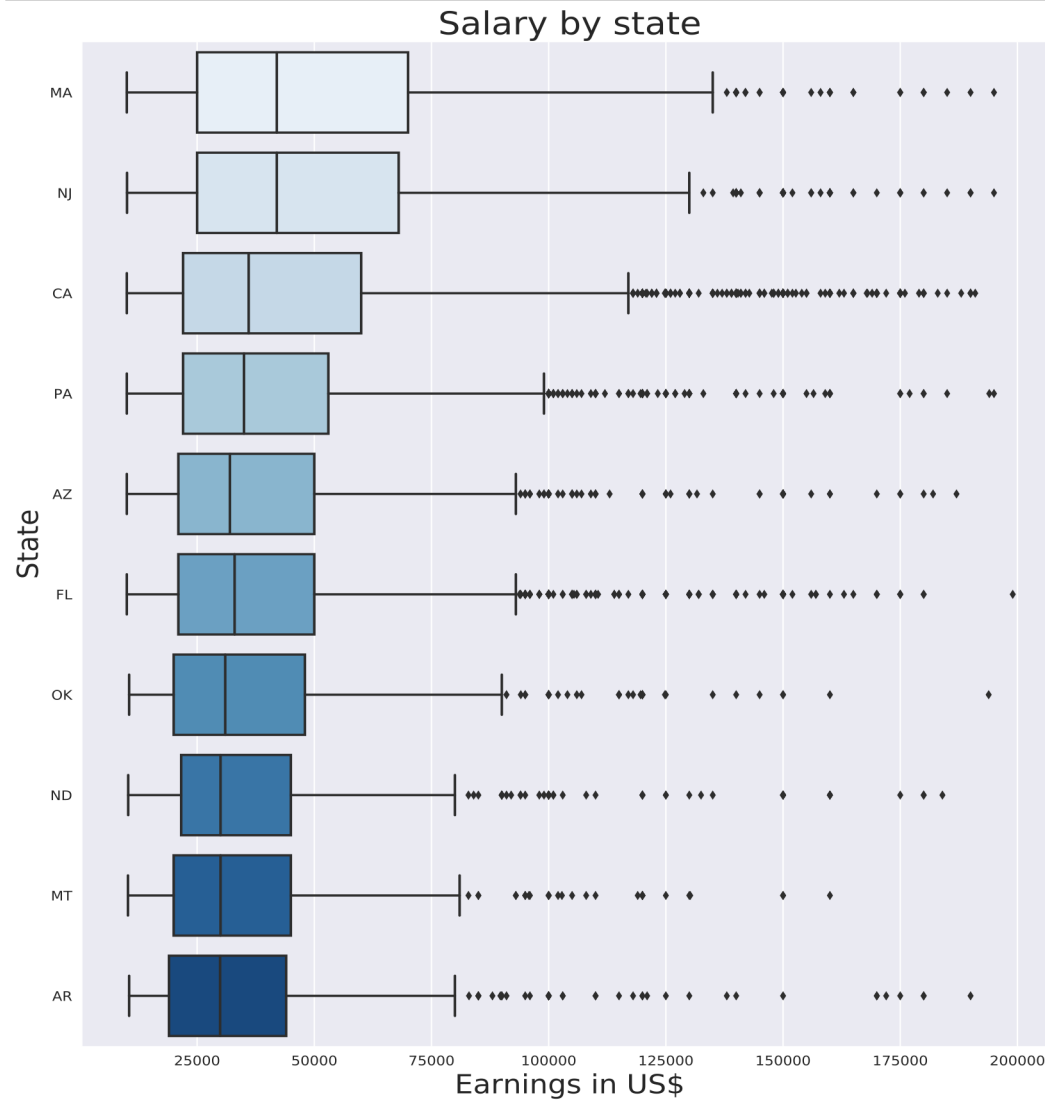
What affects our salary?

Occupation



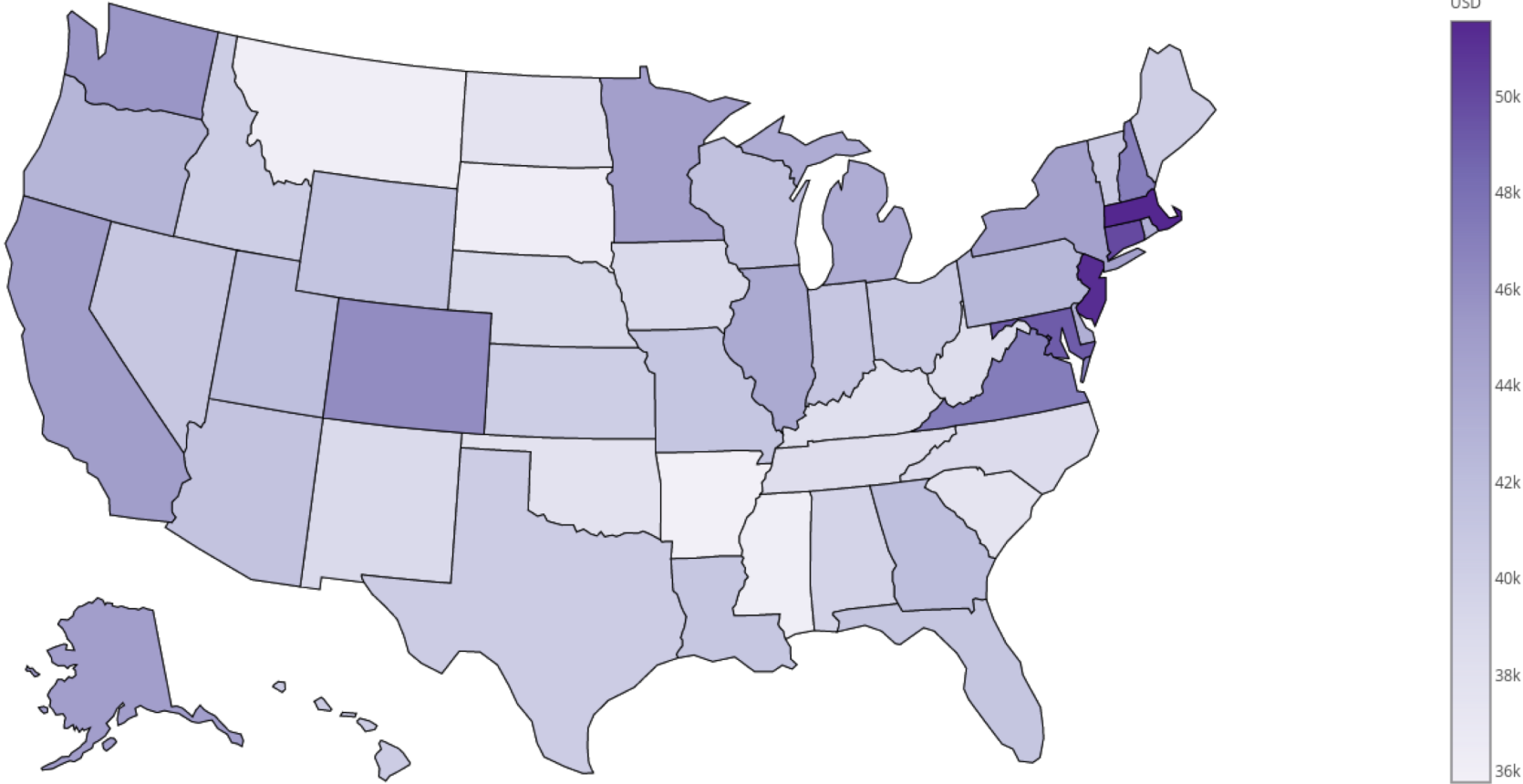
What affects our salary?

State



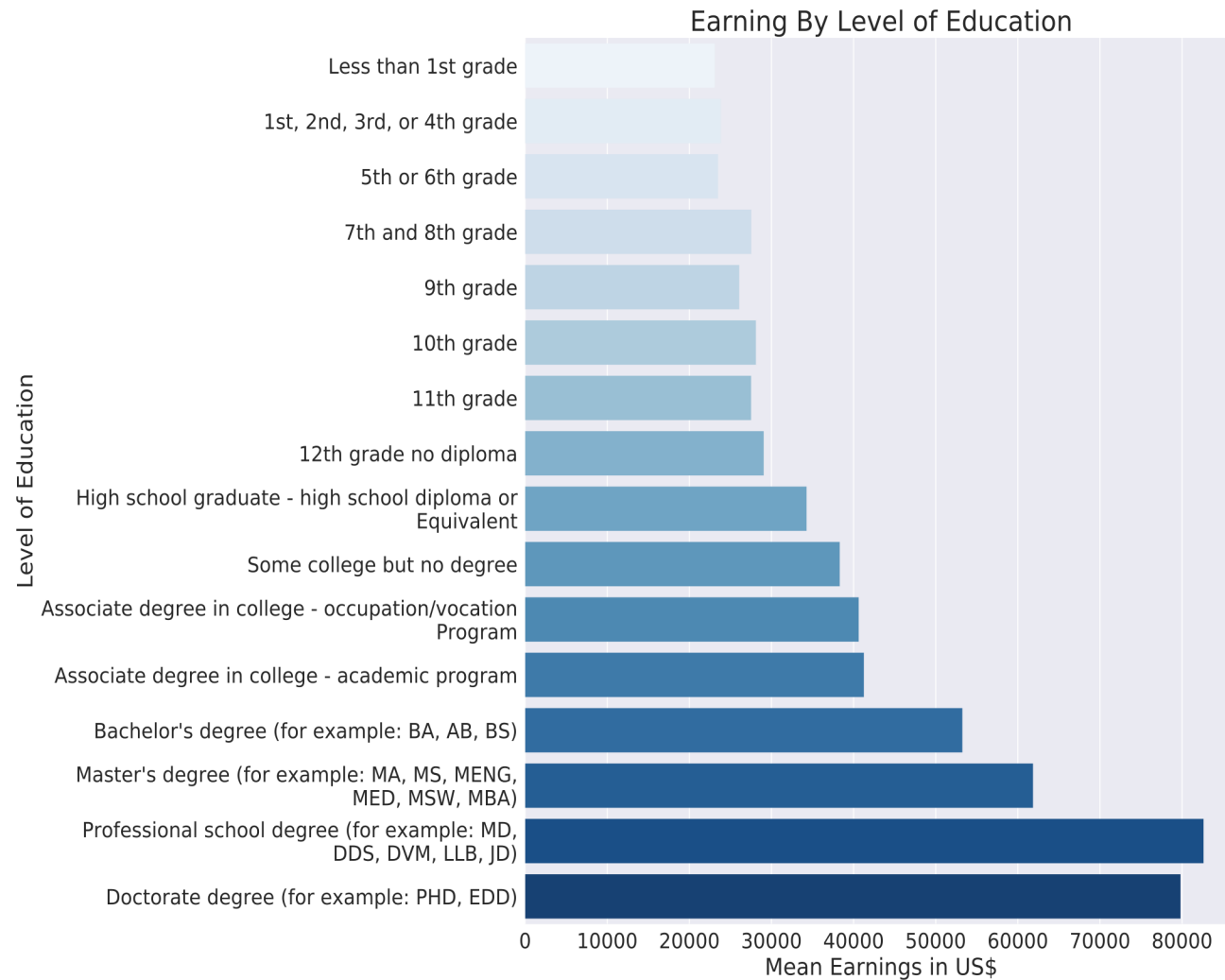
- ▶ Different states have different mean earnings
- ▶ *There is also a difference in the outliers distribution*
- ▶ *Statistically significant difference in the mean between states.*

2007 USA Mean Income By State
Population Mean = \$41.95k
STD = 3.92k



What affects our salary?

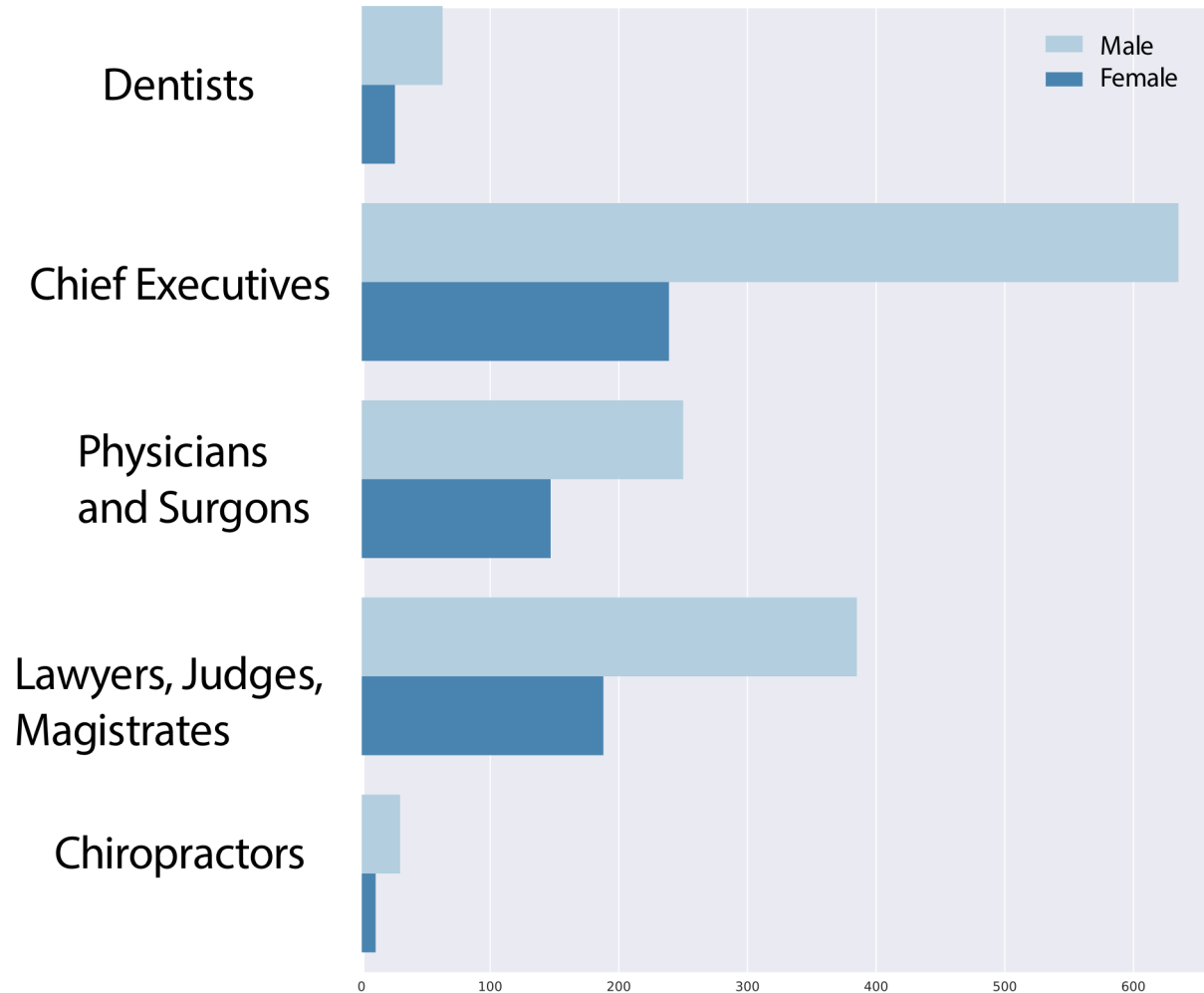
Education



- ▶ Education level is correlated with mean income.
- ▶ *There are distinct jumps for high-school graduates, college graduates and post-graduates*
- ▶ *Statistically significant difference in the mean college graduates and non graduates.*

What affects our salary?

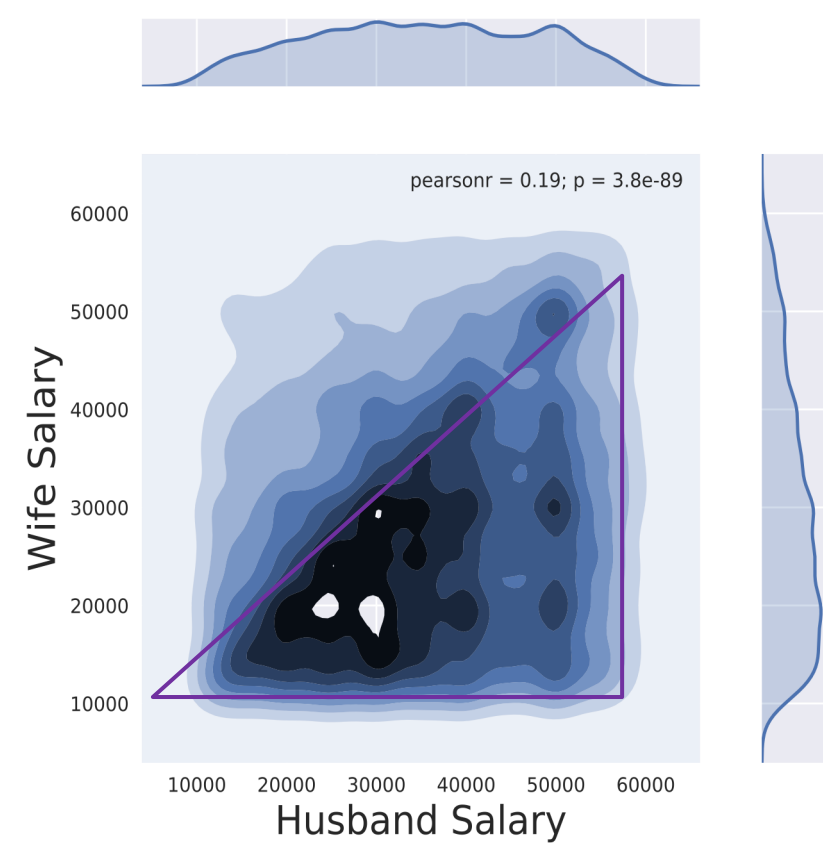
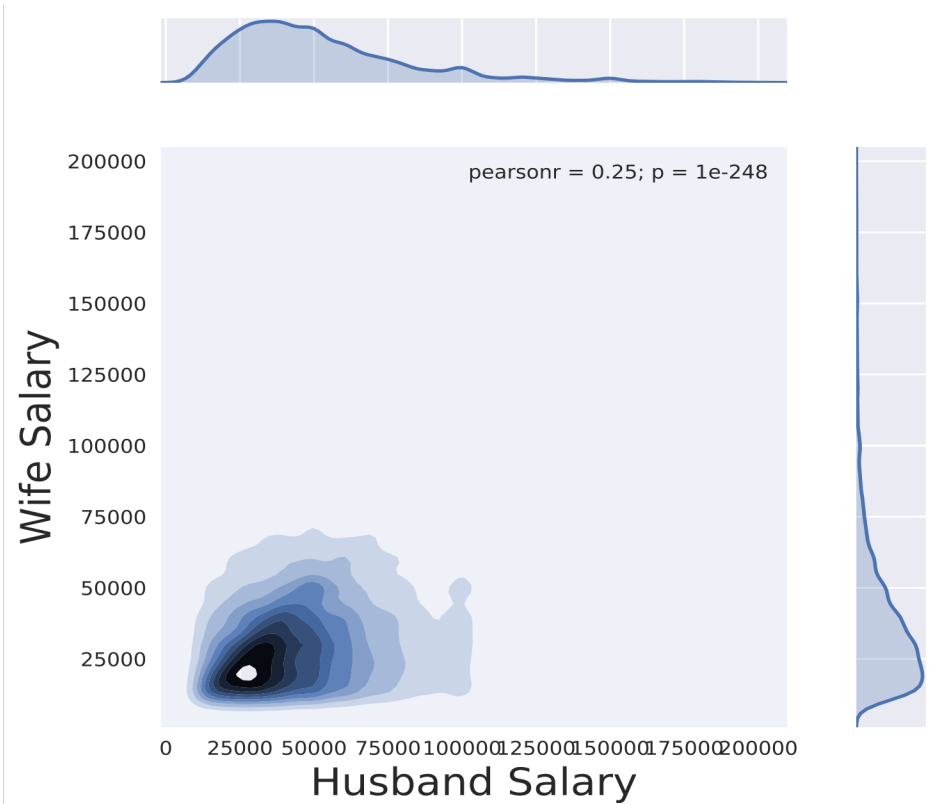
Gender



- ▶ Women are under represented in high paying occupations
- ▶ *There is also a difference in pay scale within the same occupation (not general for all occupations)*
- ▶ *Women tends to earn less then their spouses within the same household*

What affects our salary?

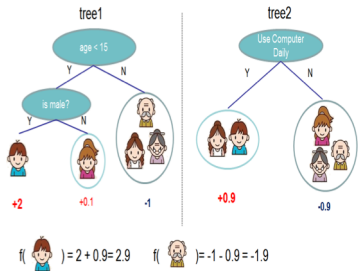
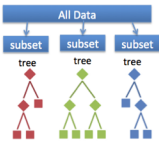
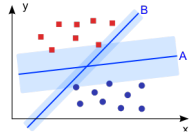
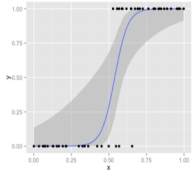
Household correlations



Classification

Comparing classifiers

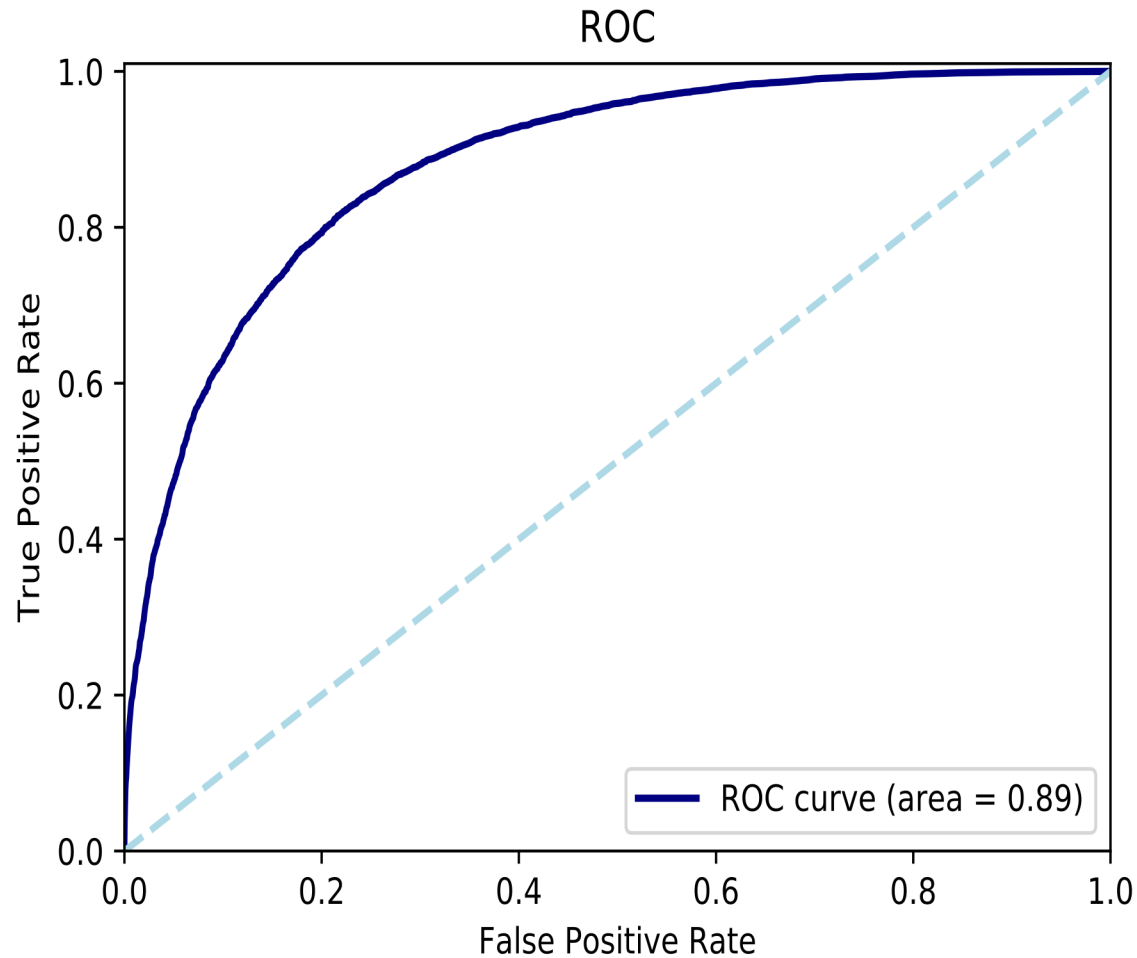
Classifier	Accuracy	Cohen's Kappa	F1-Score
Logistic Regression	0.78	0.55	0.722
SVM	0.8	0.55	0.719
Random Forest	0.81	0.56	0.71
Gradient-Boosting	0.81	0.55	0.719
XGBoost	0.81	0.58	0.724
lightGBM	0.81	0.56	0.74



- ▶ Logistic regression is fast and give good f-1 score
- ▶ SVM is has higher accuracy but worse f-1 score (“accuracy paradox”)
- ▶ For random forest we get better kappa score, but worse f-1 score, better true negative detection
- ▶ Gradient boosting performs really well. XGBoost is best but LightGBM gives similar prediction levels and is much faster

Classification

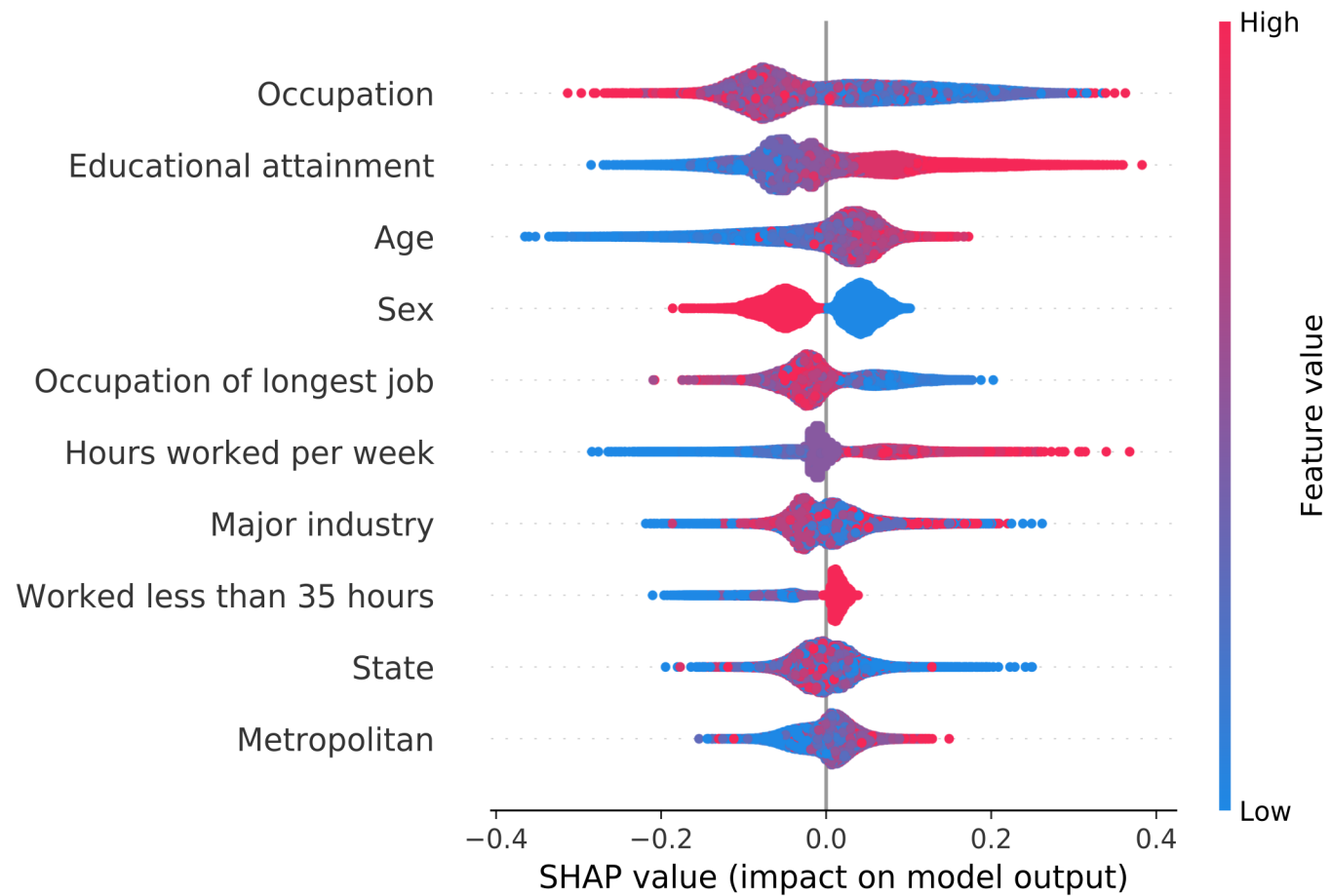
ROC curve and AUC



- ▶ ROC curve for LightGBM
- ▶ *The AUC is 0.89*
- ▶ *The threshold should be tuned according to the importance of the true positive rate*

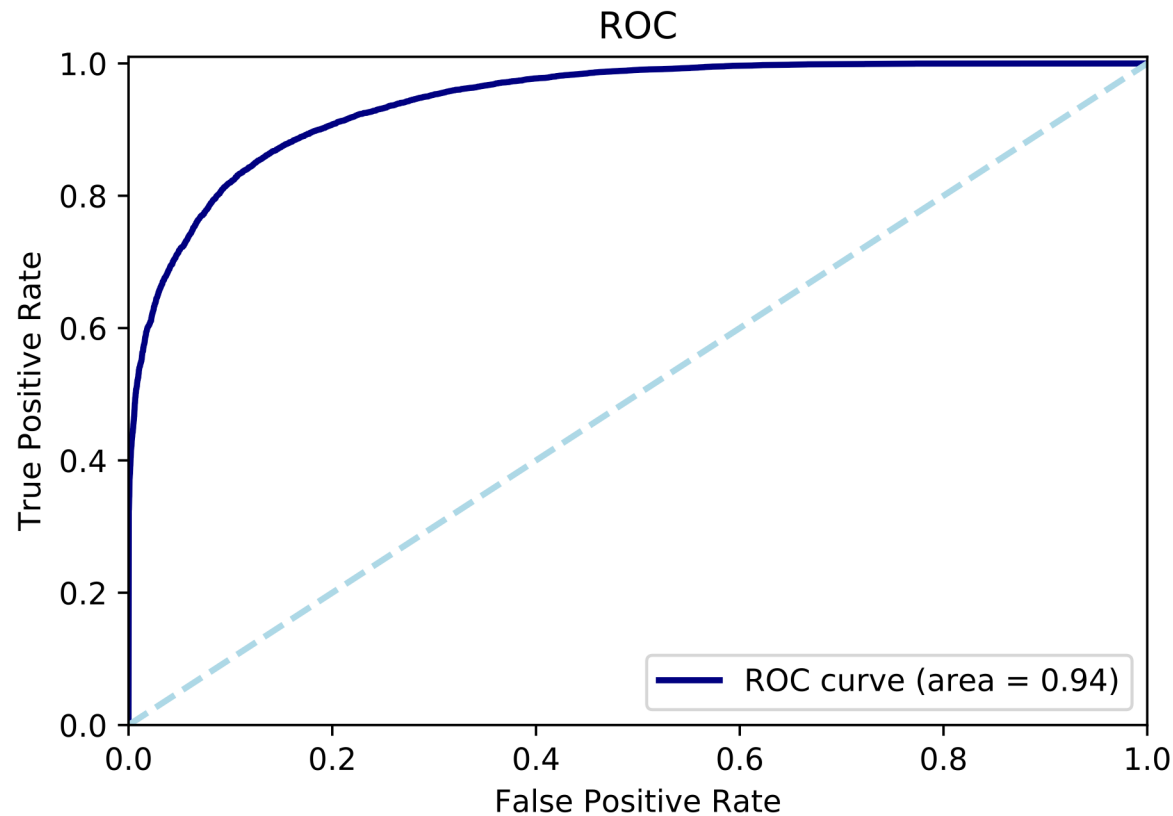
Classification

Feature importance



Classification

Predicting gender



- ▶ The classifier can be used to accurately predict gender
- ▶ $f-1 : 0.85$
- ▶ Cohen's kappa : 0.72
- ▶ Accuracy : 0.86
- ▶ Discriminative job market

Suggestions for future improvements

- ▶ Some of the occupations are underrepresented. Getting a more accurate distribution of salary for each occupation should improve the score.
- ▶ Grouping together similar occupations (or occupations with similar wage distribution)
- ▶ Sample microdata for the US Census are readily available online and contains millions of records
- ▶ Scarping Glassdor™ for income distributions
- ▶ Multi-Class classification methods