

# Predict the population of high earnings with correlated features

Team: Error 404 Sleep Not Found

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# Outline

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# Introduction & Motivation

1. Socioeconomic status increasingly plays important roles in analyzing the behaviour of society, e.g. health, incomes and educations.
2. The gap of earnings could be a potential issue.
3. We could provide some remarkable suggestions in a quantitative way to reduce the gap between the rich and poor.

# Methodology

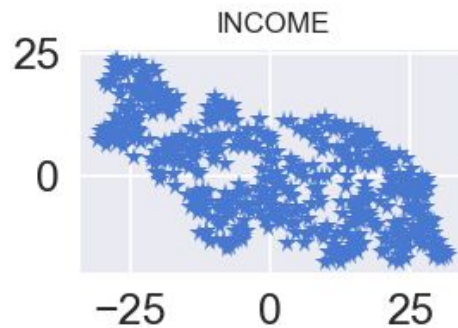
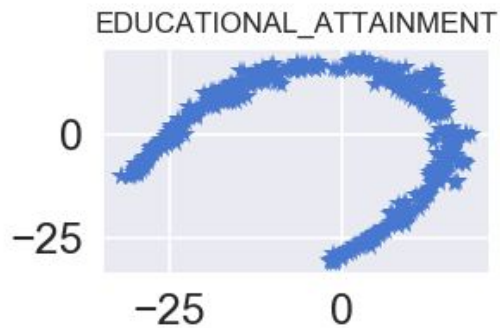
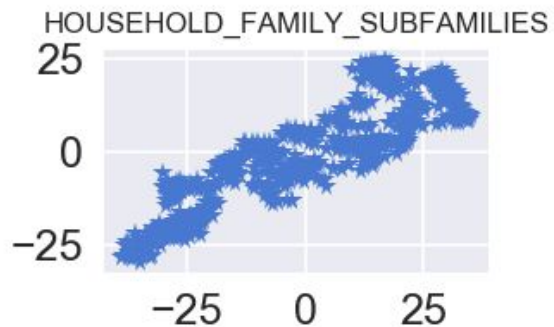
## - Feature Selection

- Calculate the Pearson correlation coefficient between the feature of the size of population of high earnings and other features
- $\rho_{X,Y} = \frac{\text{cov}(X,Y)}{\sigma_X \sigma_Y} \in [-1, +1]$
- Threshold of correlation coefficient 0.5, 0.6 and 0.7
- Separate the original dataset to training(80%) and validation data(20%)
- Remove “margin to error” features and columns with null values

## - Prediction

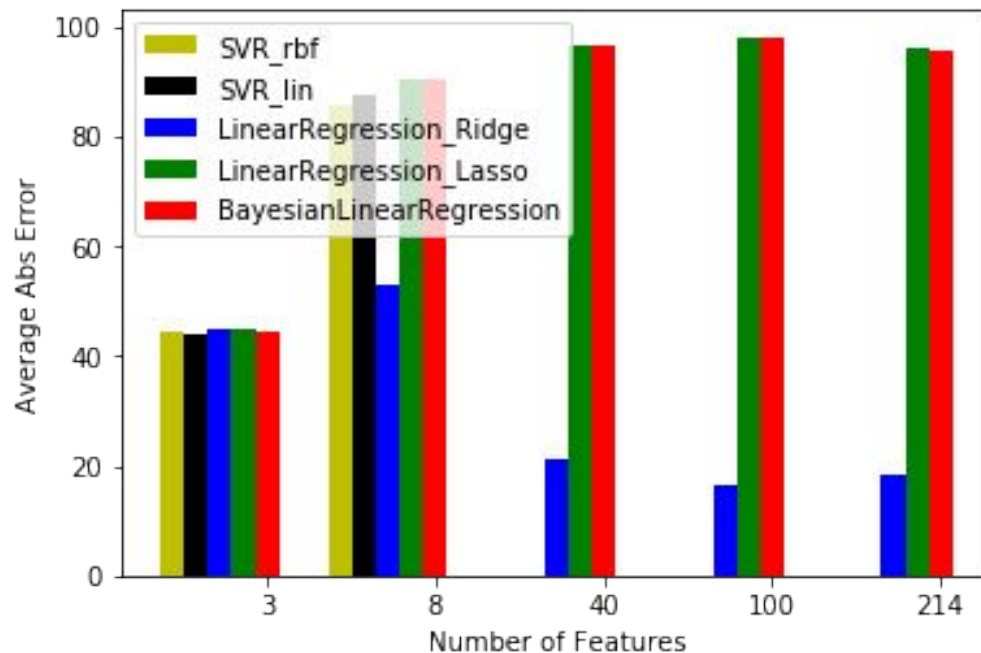
- Support vector regression
- Linear regression
- Bayesian linear regression
- etc.

# Dimensionality Reduction



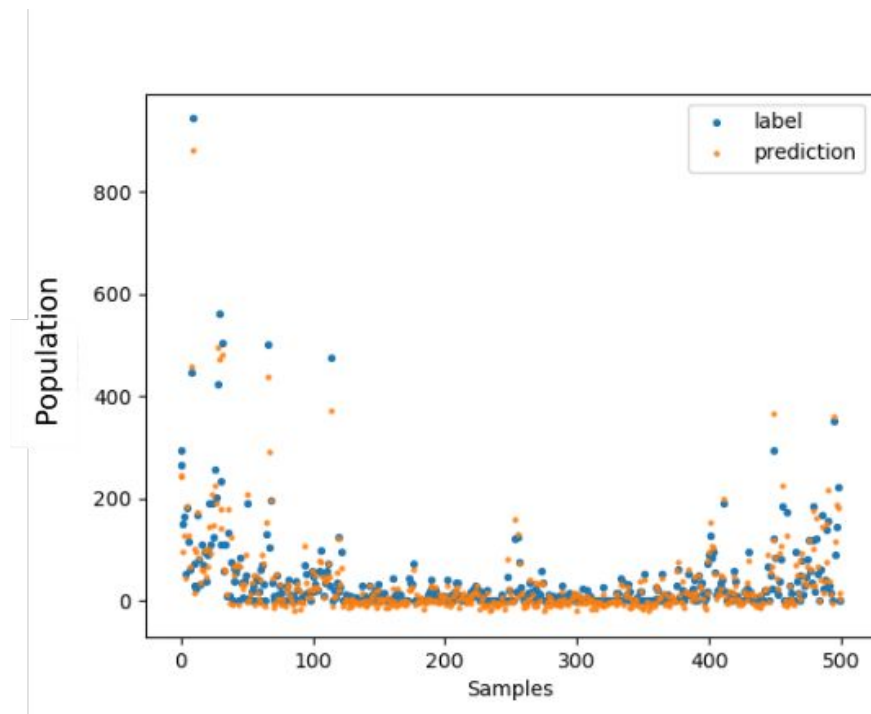
Methodology: T-distributed Stochastic Neighbor Embedding

# Results



Comparisons of different regression models with different number of features. Bayesian Linear Regression, Linear Regression with Ridge or Lasso regularization, Support Vector Regression with Linear or RBF kernel.

# Results



Prediction results are close to the true label values. ( ~ average absolute error of 16.5 for each sample, range of label value is [0,1600])

# Conclusion

- High-earning population in certain location is highly correlated to education attainment and household family relationship
- The other key features also include: poverty, income and earnings(of females in the family)
- Decent prediction results on validation set.
- Future investigation on predictions on other feature (e.g. health condition) or more robust model structure.



# Reference

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**Question?**