

ML project

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data

- Census survey
- 40 demographic features

objectives

- Predict income: more or less than \$50k
- Customer segmentation

raw data

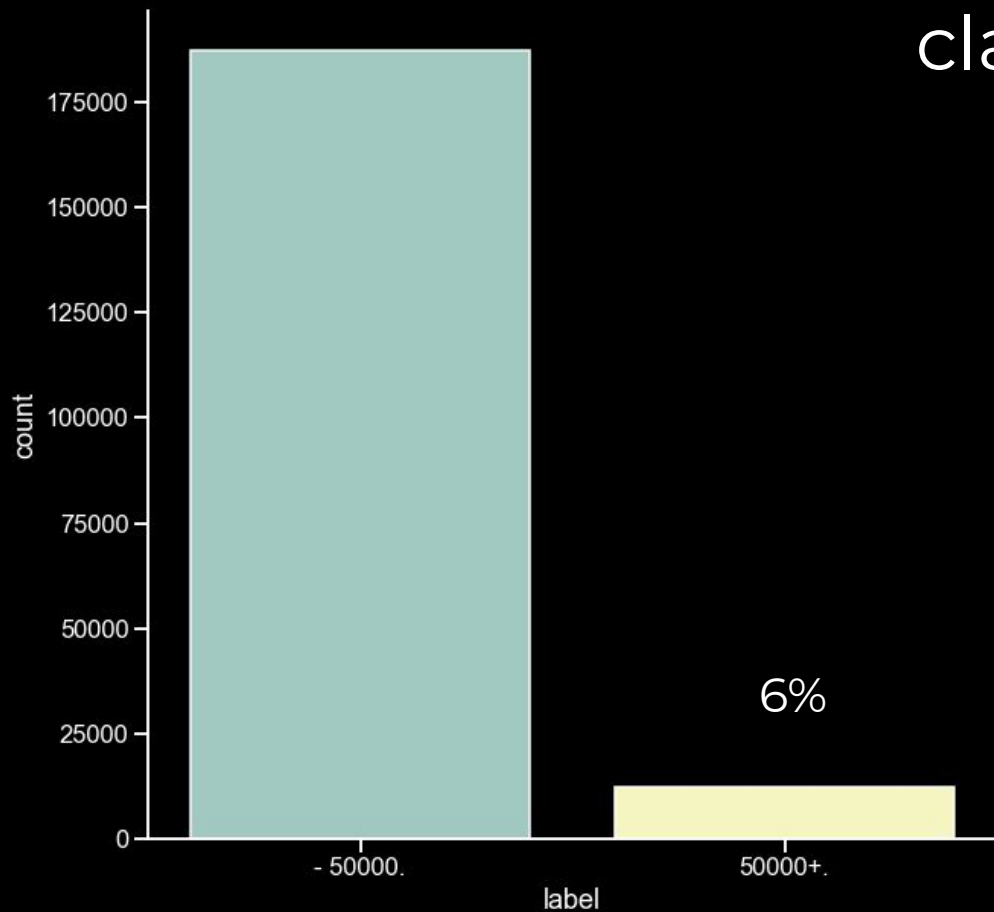
- 200k samples
- 40+2 features
 - numerical and categorical

Two are special:

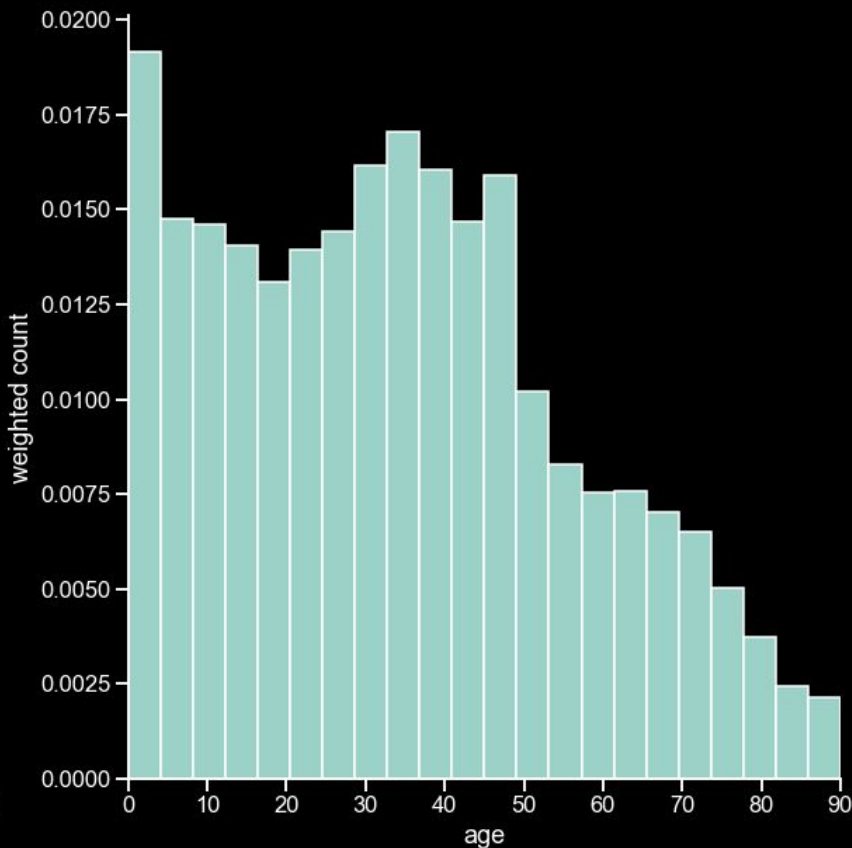
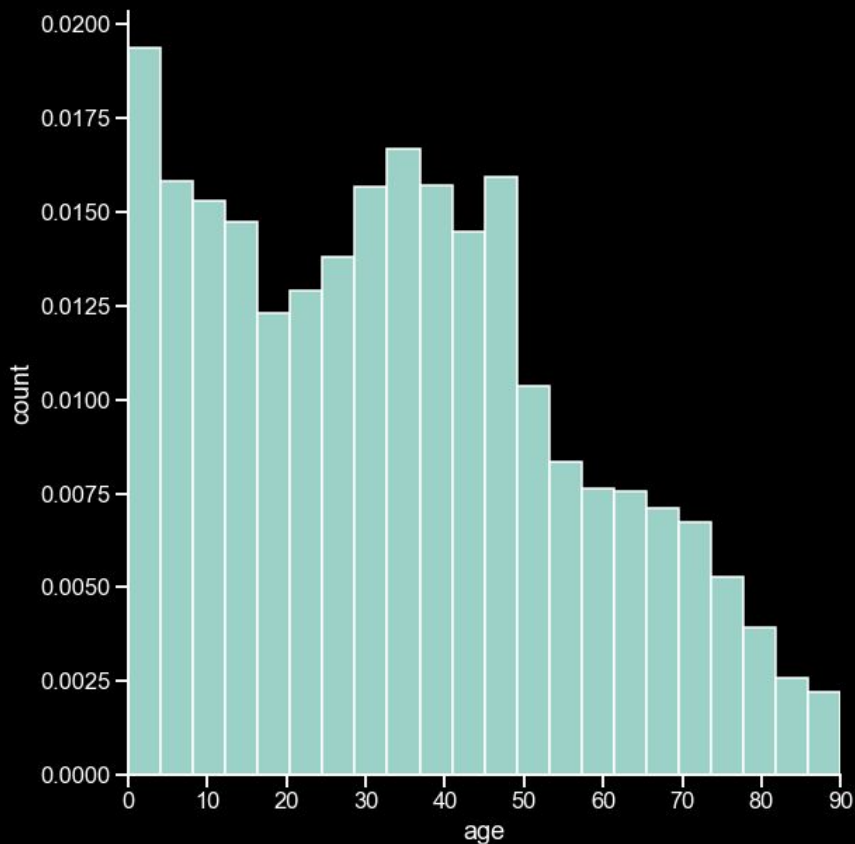
- 'label'
- 'instance weight'

| | |
|--|---------|
| age | int64 |
| class of worker | object |
| detailed industry recode | int64 |
| detailed occupation recode | int64 |
| education | object |
| wage per hour | int64 |
| enroll in edu inst last wk | object |
| marital stat | object |
| major industry code | object |
| major occupation code | object |
| race | object |
| hispanic origin | object |
| sex | object |
| member of a labor union | object |
| reason for unemployment | object |
| full or part time employment stat | object |
| capital gains | int64 |
| capital losses | int64 |
| dividends from stocks | int64 |
| tax filer stat | object |
| region of previous residence | object |
| state of previous residence | object |
| detailed household and family stat | object |
| detailed household summary in household | object |
| instance weight | float64 |
| migration code-change in msa | object |
| migration code-change in reg | object |
| migration code-move within reg | object |
| live in this house 1 year ago | object |
| migration prev res in sunbelt | object |
| num persons worked for employer | int64 |
| family members under 18 | object |
| country of birth father | object |
| country of birth mother | object |
| country of birth self | object |
| citizenship | object |
| own business or self employed | int64 |
| fill inc questionnaire for veteran's admin | object |
| veterans benefits | int64 |
| weeks worked in year | int64 |
| year | int64 |
| label | object |
| dtype: object | |

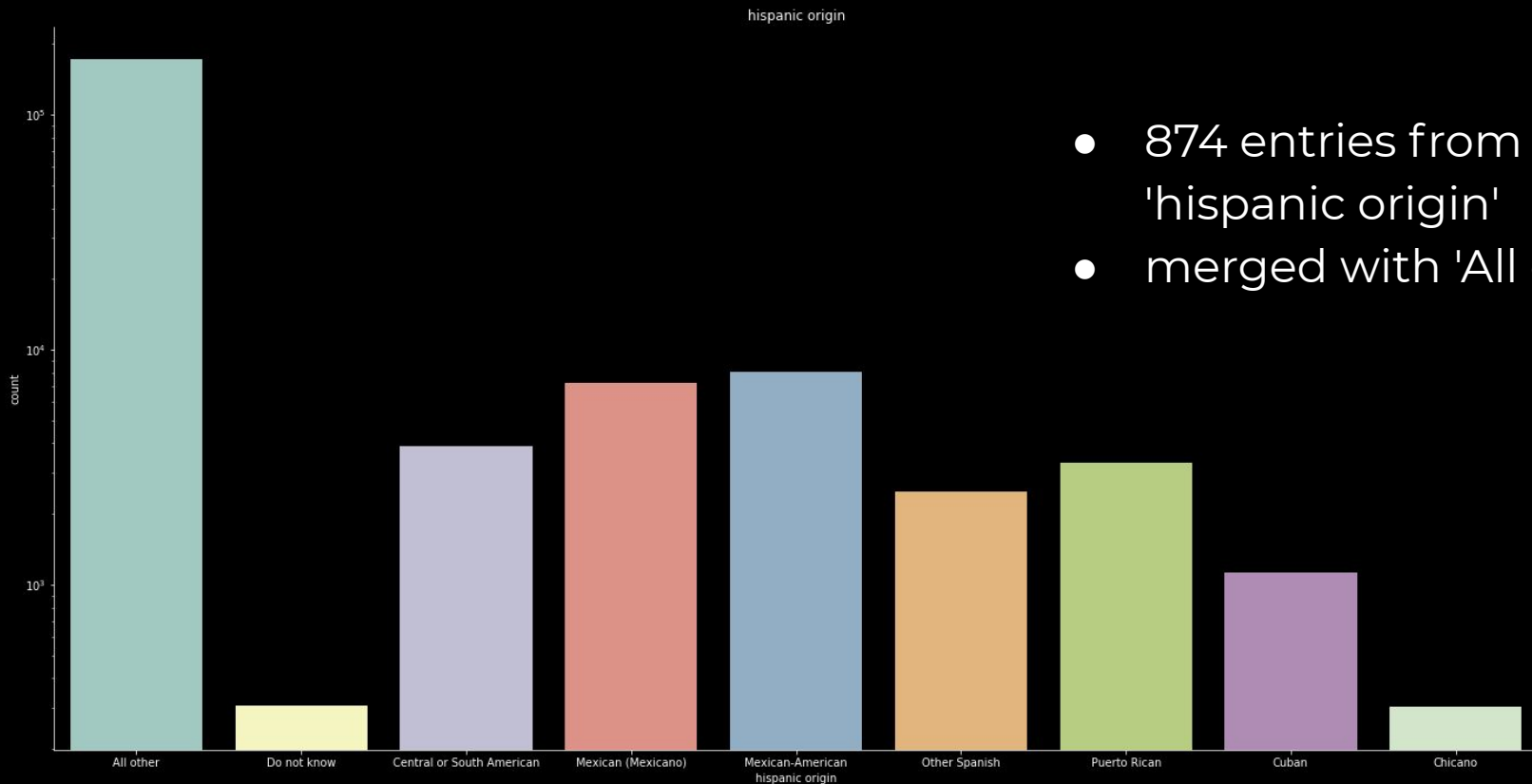
class imbalance



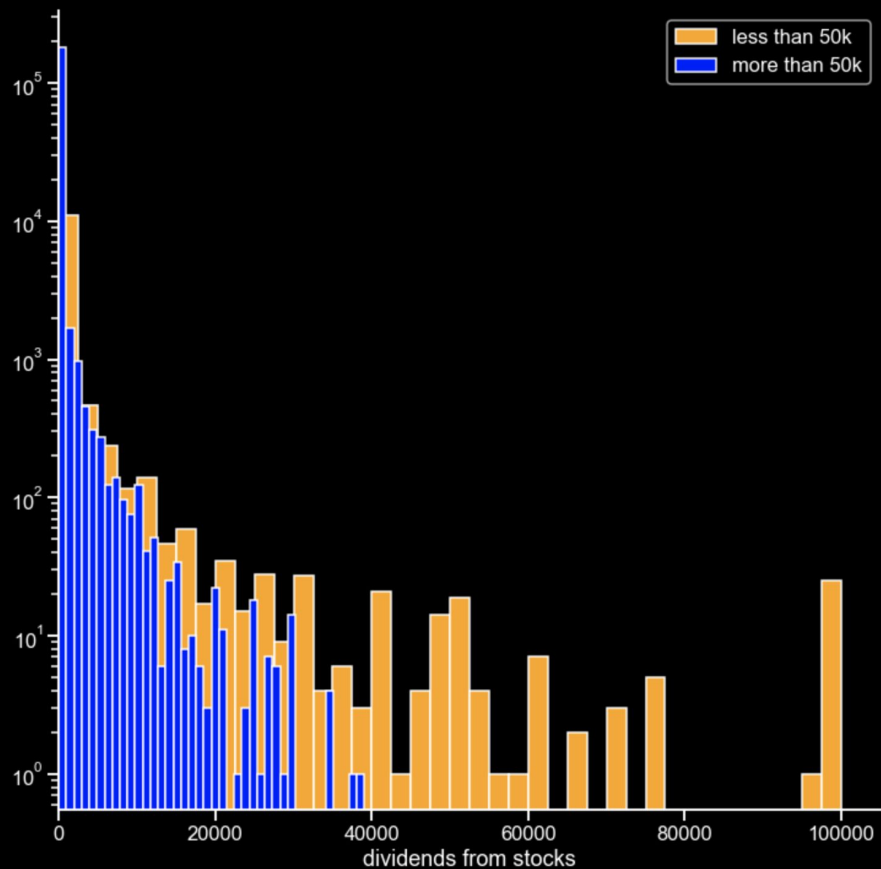
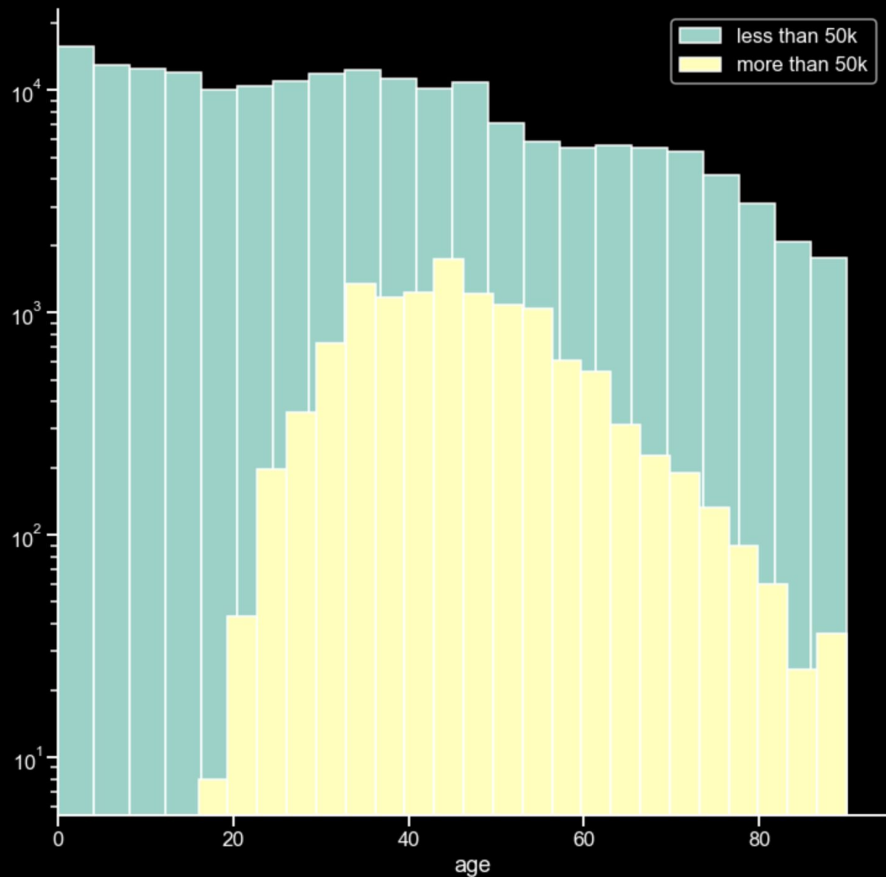
instance weight (sanity check)



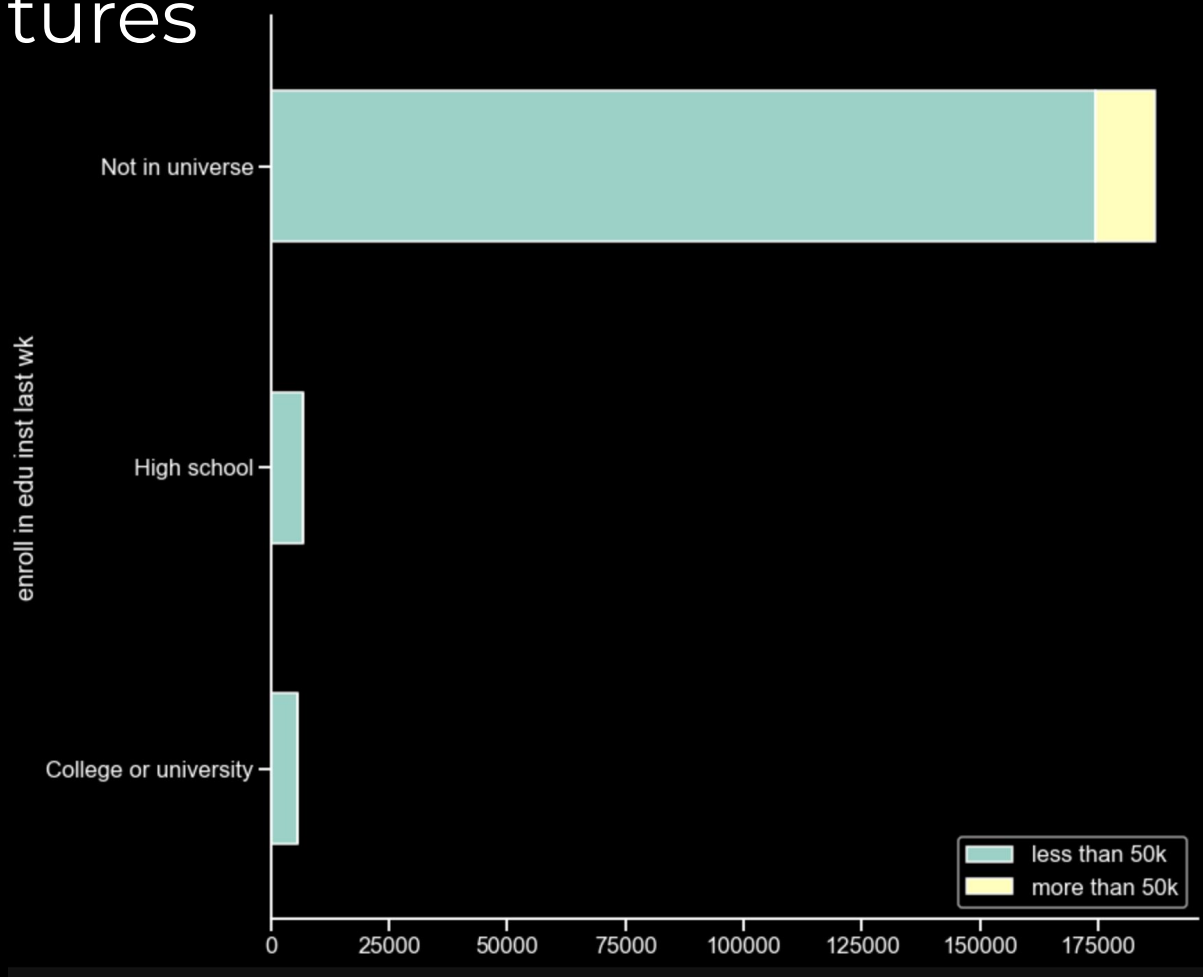
missing values



numerical features

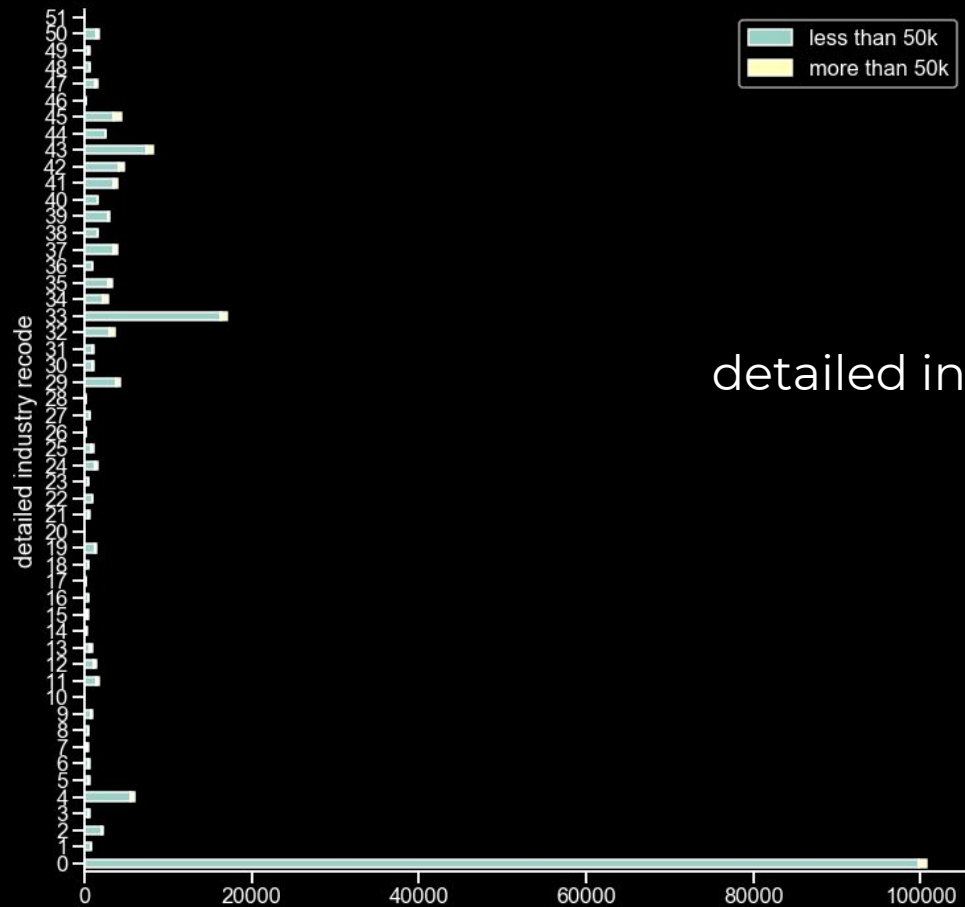


categorical features



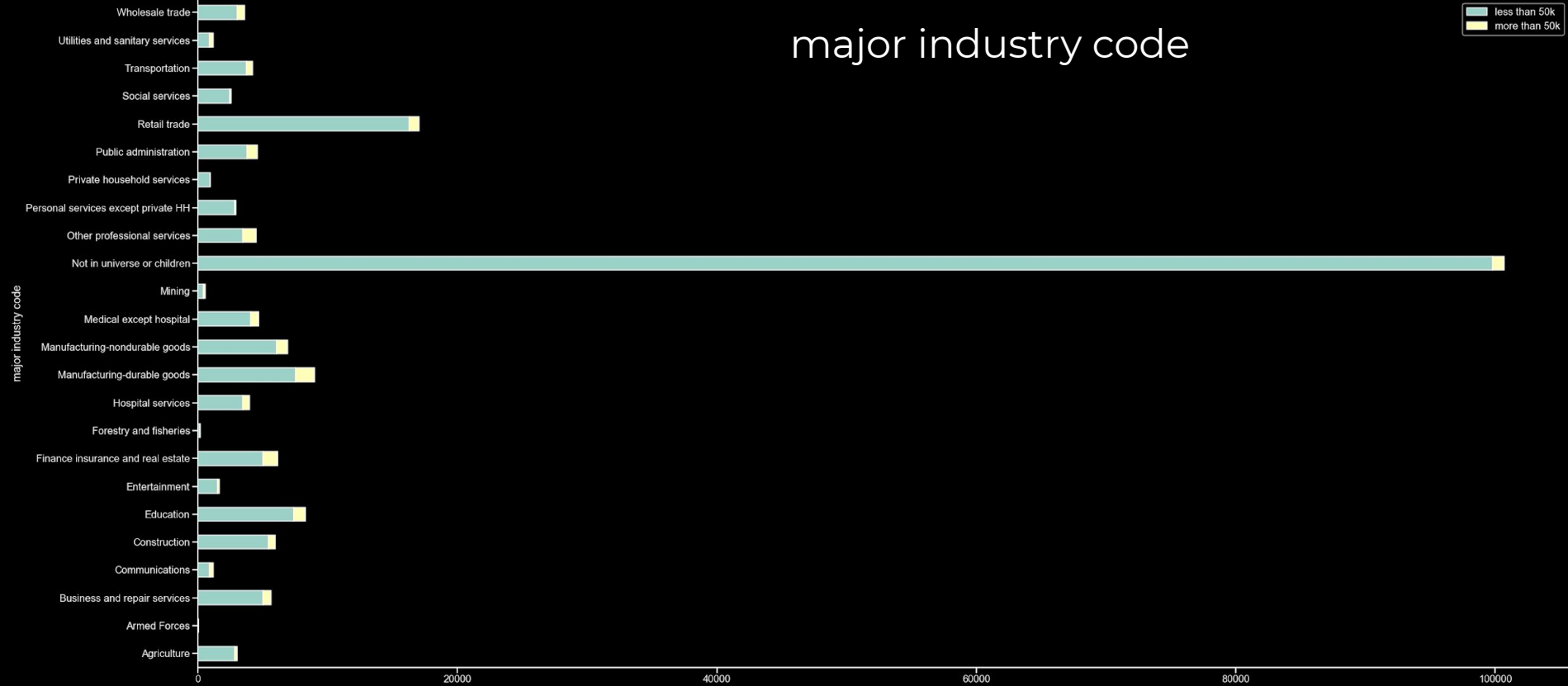
cleaning

- 1/0 encoding of binary features (e.g. label, sex, year)
- one-hot encoding of categorical features
- drop duplicate features (high linear correlation)



detailed industry recode

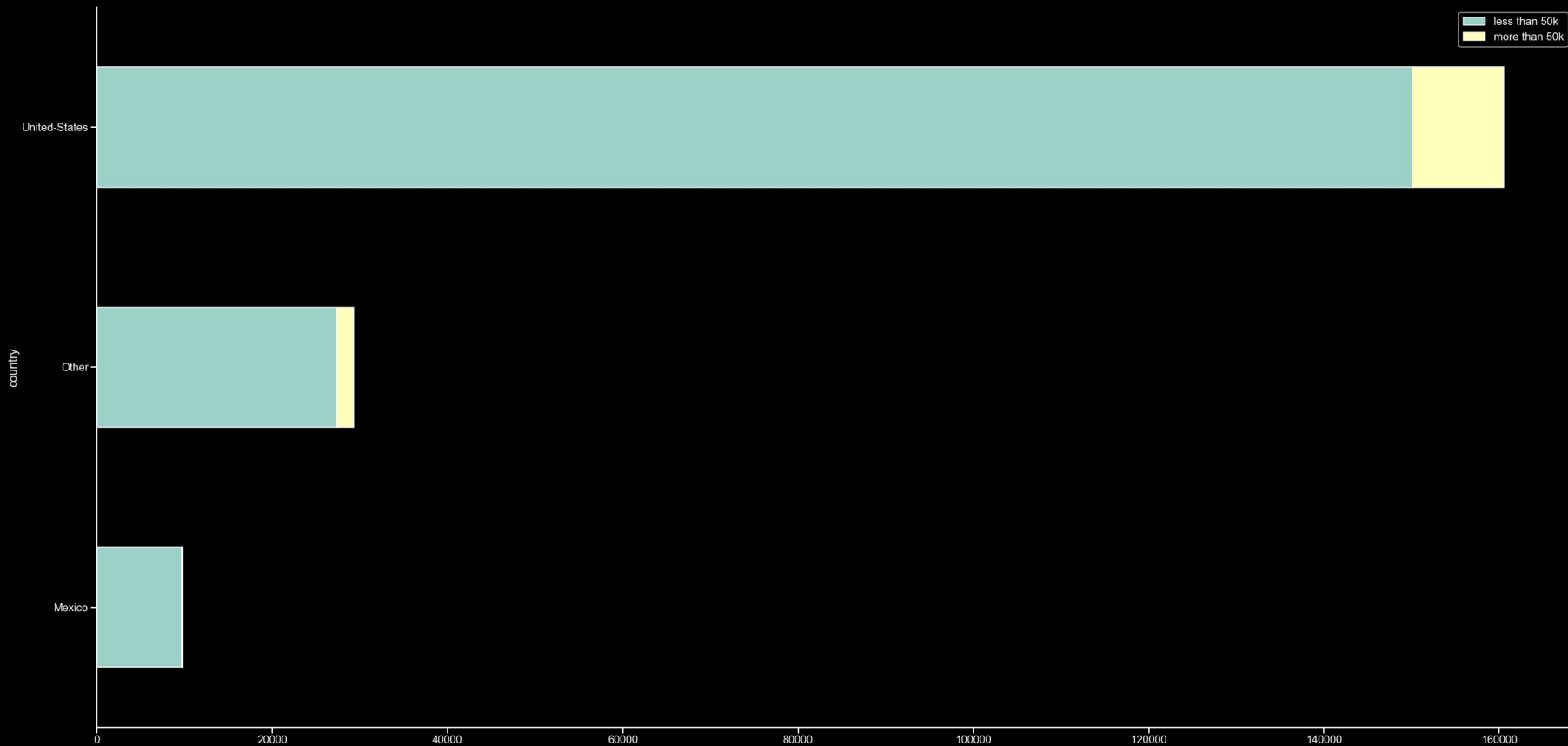
major industry code



more cleaning



more cleaning



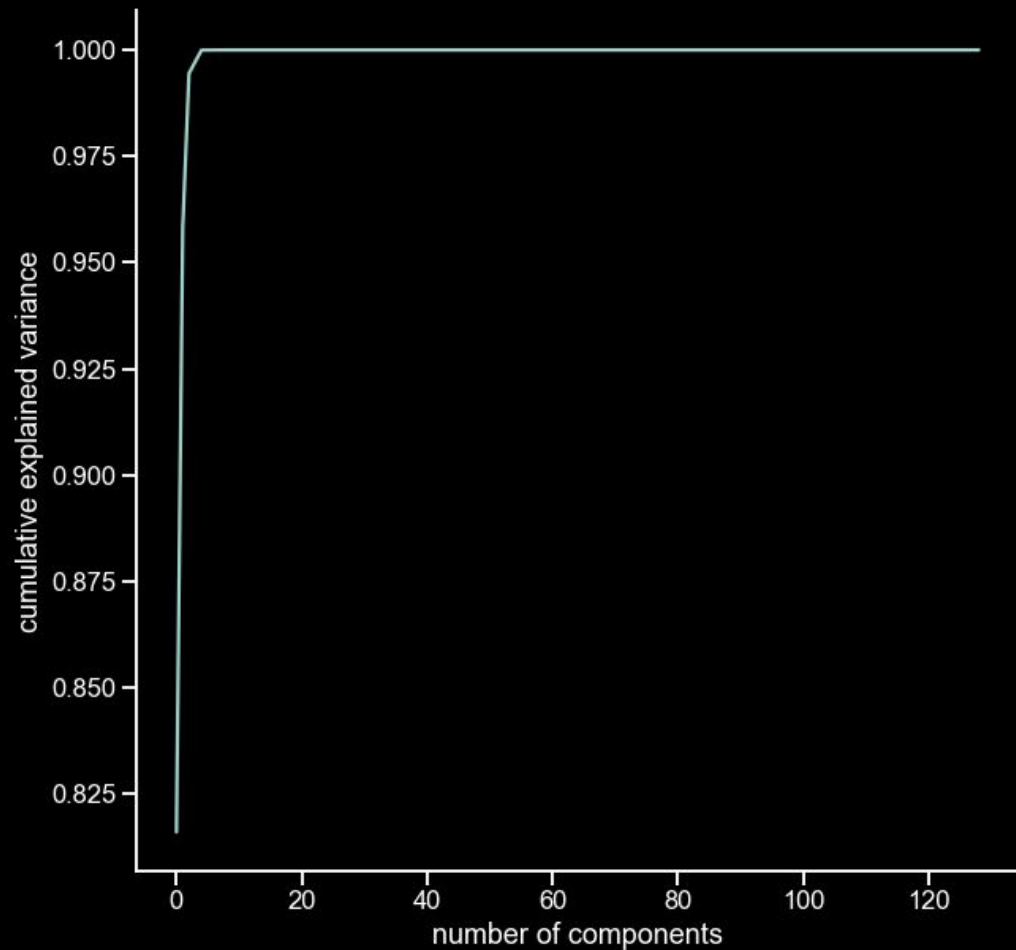
cleaned data

- 26+2 features
(before one-hot
encoding)

| | |
|---|---------|
| age | int64 |
| class of worker | object |
| education | object |
| wage per hour | int64 |
| enroll in edu inst last wk | object |
| marital stat | object |
| major occupation code | object |
| race | object |
| hispanic origin | object |
| member of a labor union | object |
| reason for unemployment | object |
| full or part time employment stat | object |
| capital gains | int64 |
| capital losses | int64 |
| dividends from stocks | int64 |
| tax filer stat | object |
| region of previous residence | object |
| detailed household summary in household | object |
| instance weight | float64 |
| num persons worked for employer | int64 |
| family members under 18 | object |
| own business or self employed | int64 |
| veterans benefits | int64 |
| weeks worked in year | int64 |
| label_encoded | int64 |
| sex_encoded | int64 |
| year_encoded | int64 |
| country | object |
| dtype: | object |

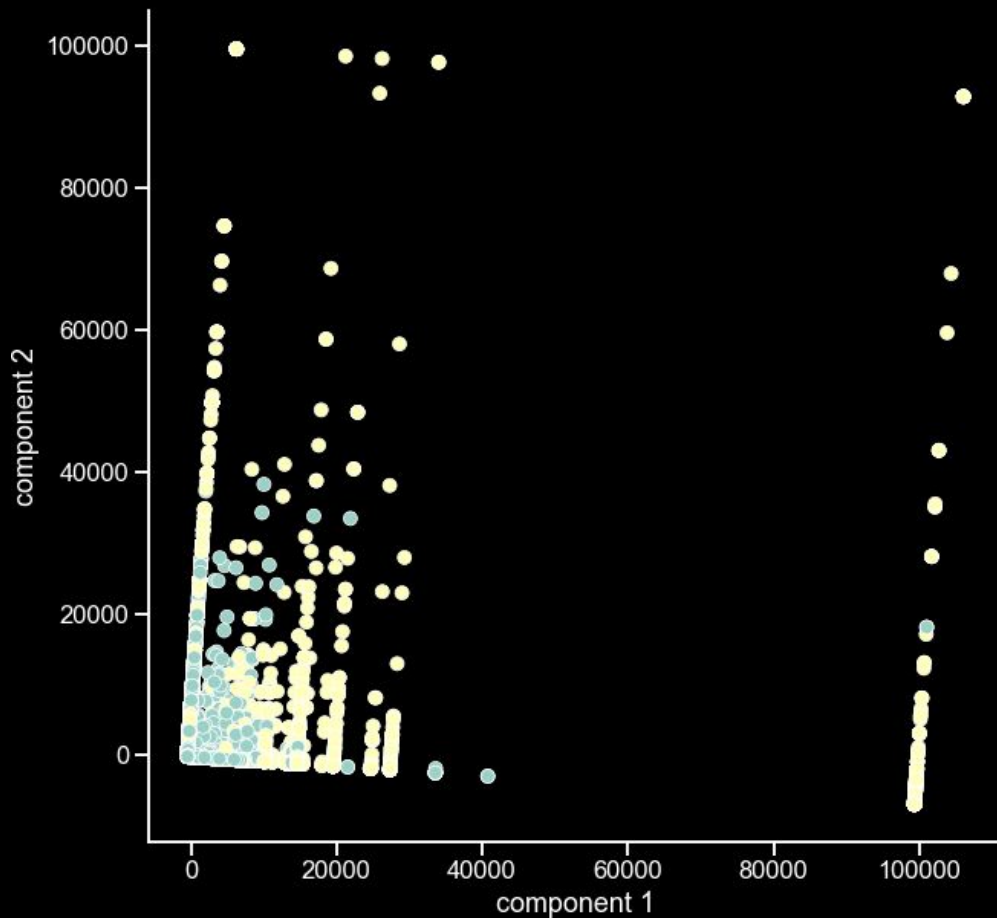
PCA

- explained variance:
2 components overwhelm



PCA: visualized

- PCA projection onto first two components
- data squashed near origin

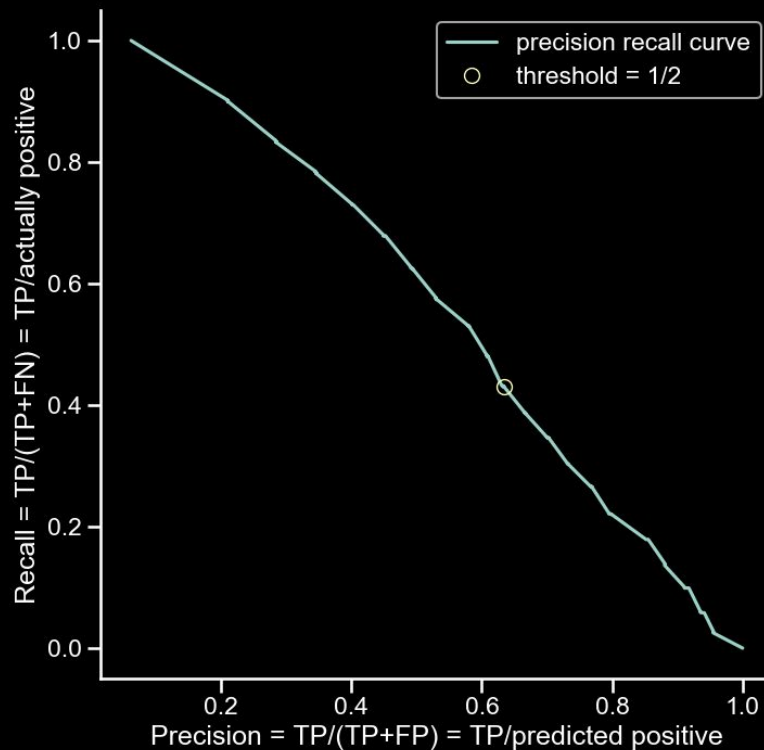
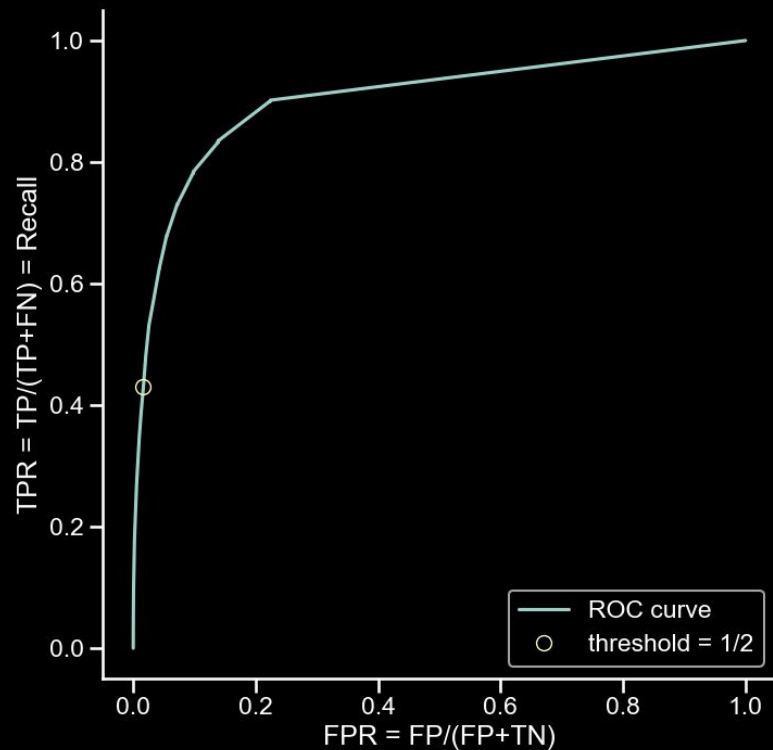


income prediction

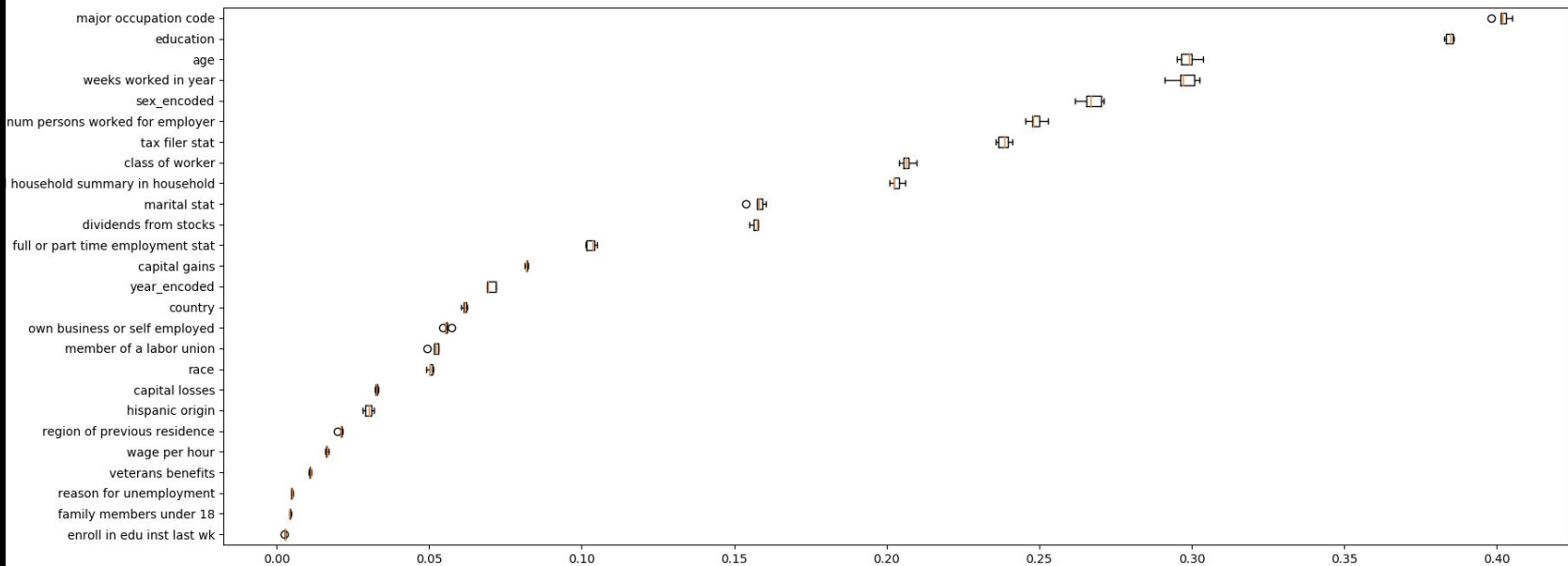
- 80/20 train/test split
- logistic regression + L1 regularization: gridsearch+CV
- random forest with 20 trees

random forest performed better

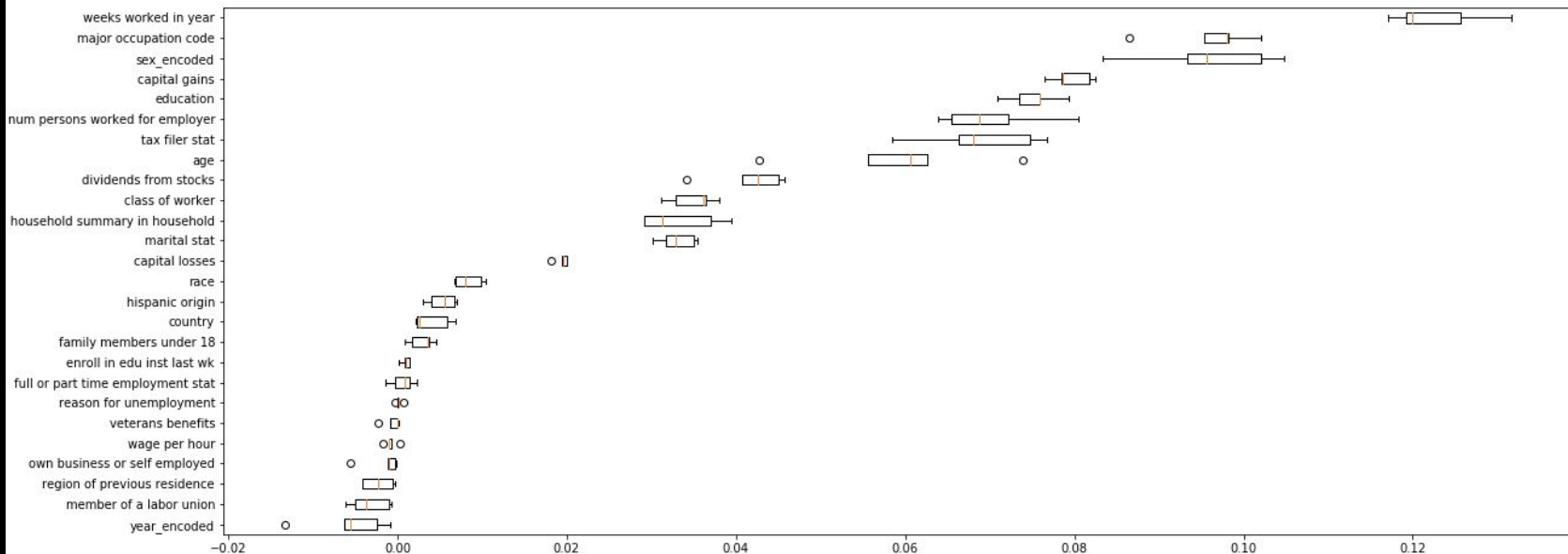
roc and precision/recall



feature importance (permutation): train



feature importance (permutation): test

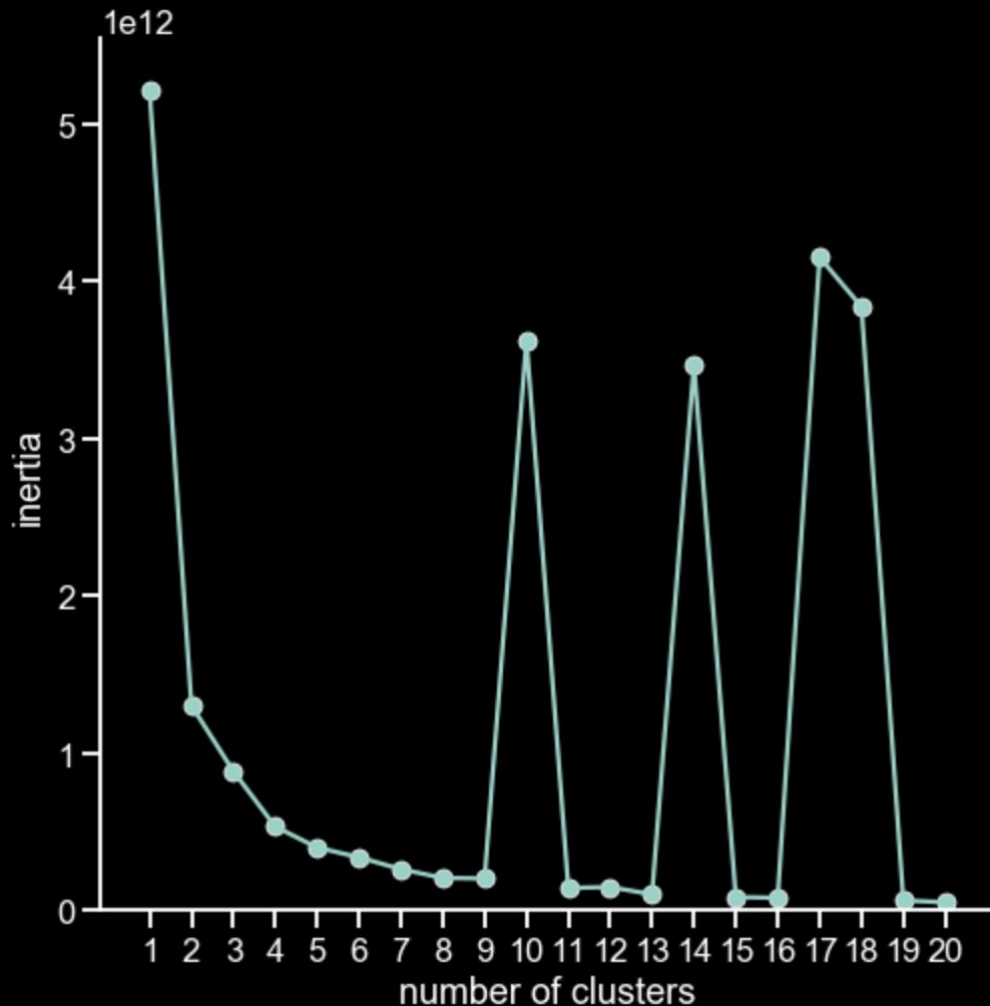


segmentation

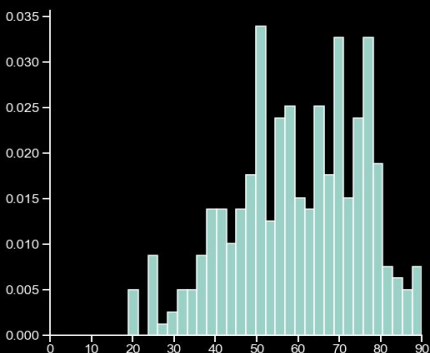
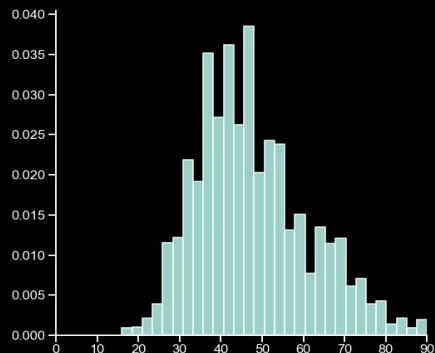
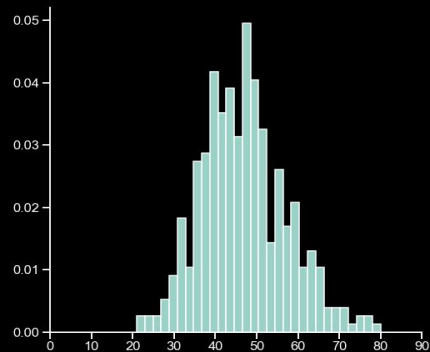
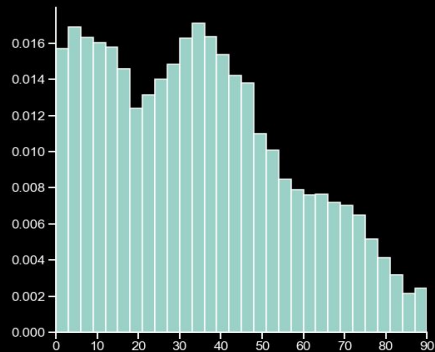
- ~~hierarchical clustering~~ (too slow)
- k-means (with mini-batches)

how many clusters?

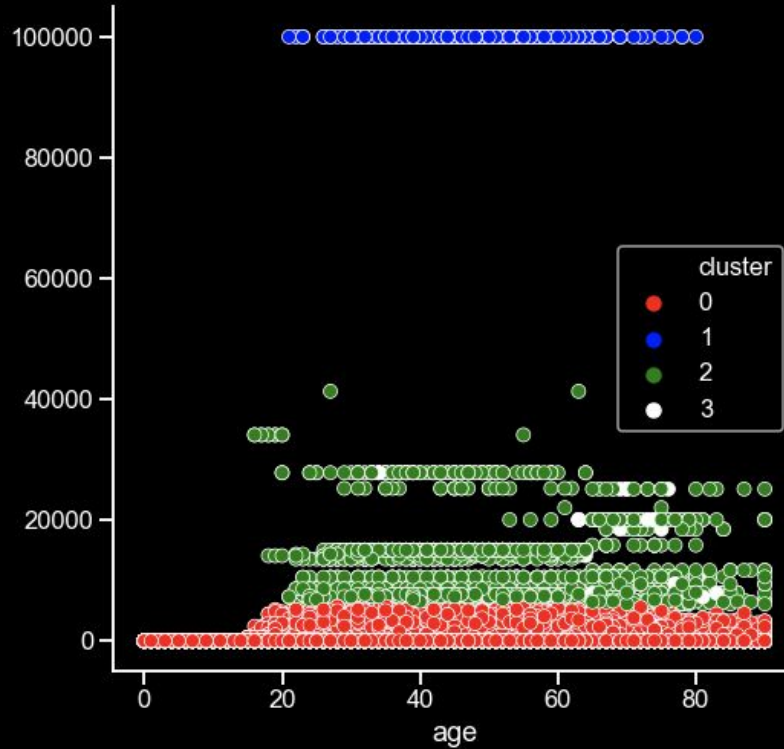
- mini-batch k-means
- batch size = 10k
 - (5% of samples)
- --> 4 clusters



what do the clusters look like?



what do the clusters look like?



next steps

- income prediction
 - fancier tree model (gradient boosted, XGBoost)
 - increase feature selection
- segmentation
 - different clustering method
 - with clear goal: rules + clustering

thanks