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Inference

```
In [1]: import pandas as pd
   import numpy as np
   import matplotlib.pyplot as plt
   %matplotlib inline
   import seaborn as sns
   sns.set_context('poster')
   import random
   from sklearn.preprocessing import StandardScaler
```

```
In [2]: from project_helper import *
```

Separate Data Subsets

```
In [3]: # read source data
df = pd.DataFrame.from_csv('../data/merged/all_data_2006_to_2016.csv', index_col=None)
```

```
In [4]: # seed for consistent splits
        random.seed(1636)
        # all idxs
        idxs = list(range(len(df)))
        # shuffle and split
        random.shuffle(idxs)
        split = int(len(idxs)*0.7)
        train all = idxs[:split]
        test all = idxs[split:]
        # idxs of outliers
        outlier idxs = list((df.MSA abbr == 'NEW ORLEANS LA').nonzero()[0])
        outlier idxs += list(((df.MSA abbr == 'MEMPHIS TN') & (df.year == 2016)).nonzero()[0])
        outlier idxs += list(((df.MSA abbr == 'BATON ROUGE LA') & (df.year == 2007)).nonzero()[0])
        # cleaned idxs
        train idxs = [idx for idx in train all if idx not in outlier idxs]
        test idxs = [idx for idx in test all if idx not in outlier idxs]
```

Train Len: 640 Test Len: 276

```
In [6]: # drop irrelevant columns
        x features df = df.drop(['MSA orig', 'MSA corr', 'MSA abbr', 'murder per 100 k'], axis=1)
        # standardize, fitting only on training rows
        standardizer = StandardScaler().fit(x features df.iloc[train idxs])
        x features = pd.DataFrame(standardizer.transform(x features df), columns=x features df.columns)
        # train test split
        x train = x features.iloc[train idxs]
        x test = x features.iloc[test idxs]
        print('Train X and y match: {}'.format(len(x train)==len(y train)))
        print('Test X and y match: {}'.format(len(x test)==len(y test)))
        print('\nNumber of Predictors: {}'.format(len(x train.columns)))
        print('\nFEATURE NAMES:')
        for name in x train.columns:
            print(' -'+name)
        Train X and y match: True
        Test X and y match: True
        Number of Predictors: 9
        FEATURE NAMES:
         -vear
         -now married except separated
         -less than high school diploma
         -unmarried portion of women 15 to 50 years who had a birth in past 12 months
         -households with food stamp snap benefits
         -percentage married-couple family
         -percentage female householder no husband present family
         -poverty all people
         -house median value (dollars)
```

Fitting The Models

```
In [7]: from sklearn.linear_model import LinearRegression, RidgeCV, LassoCV, BayesianRidge, HuberRegressor
    from sklearn.model_selection import GridSearchCV
    from sklearn.neural_network import MLPRegressor
    from sklearn.neighbors import KNeighborsRegressor
    from sklearn.ensemble import AdaBoostRegressor
    from sklearn.svm import SVR
```

```
In [14]: # instantiate and fit models
def make_models(x_train, y_train):
    md = dict()

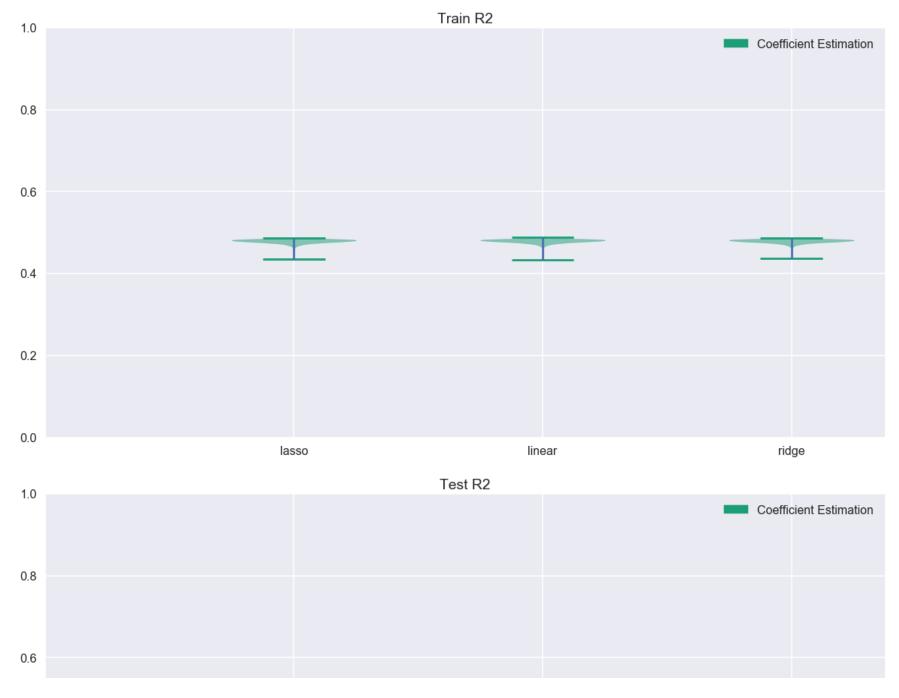
md['linear'] = LinearRegression().fit(x_train, y_train)
    md['ridge'] = RidgeCV(cv=5).fit(x_train, y_train)
    md['lasso'] = LassoCV(cv=5).fit(x_train, y_train)

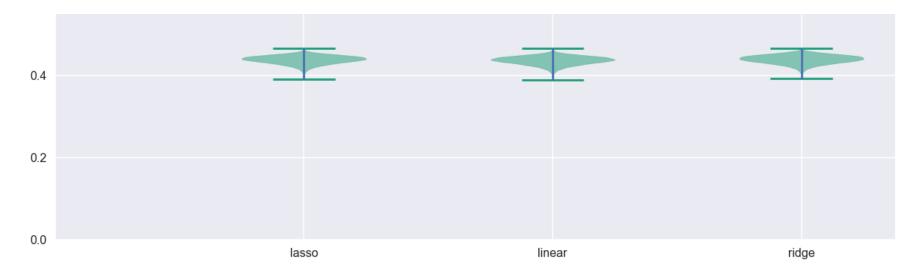
return md
```

Done

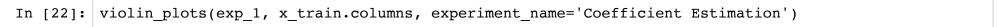
Accuracy Results

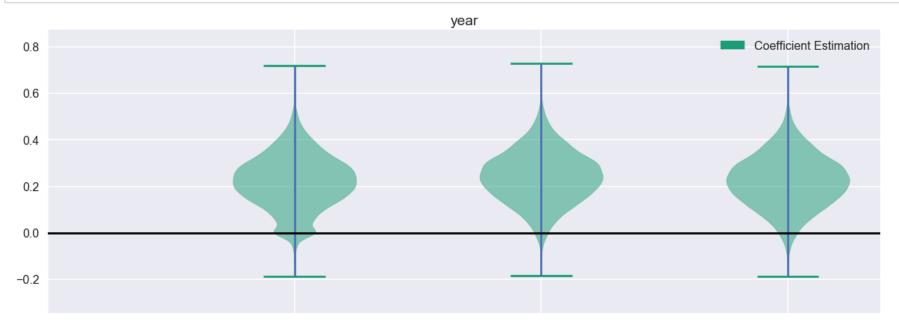
```
In [21]: violin_plots(exp_1, ['Train R2','Test R2'], experiment_name='Coefficient Estimation', center_zero=Fa
```

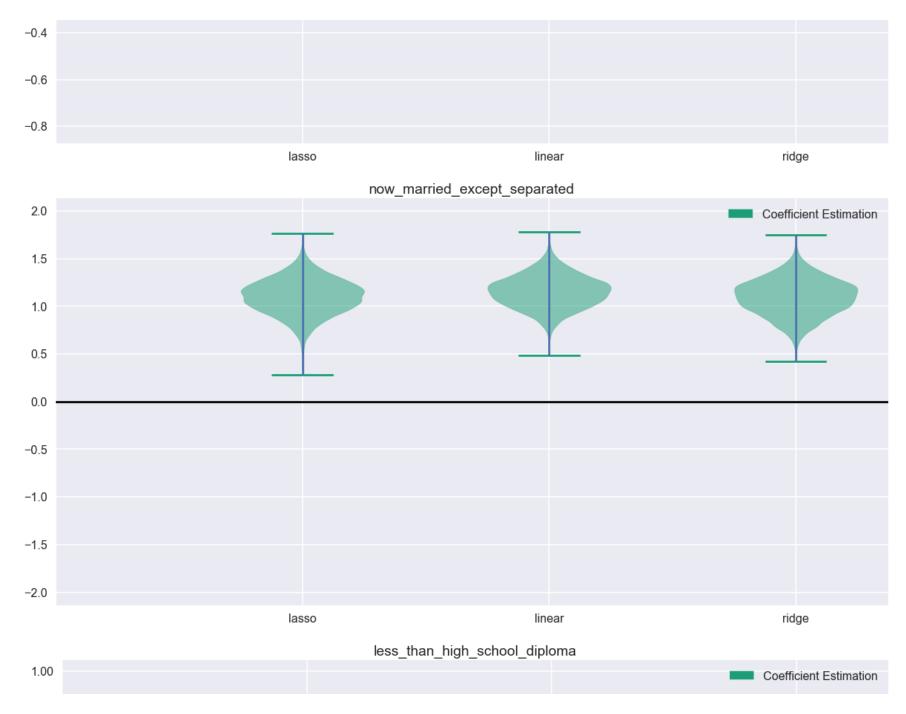


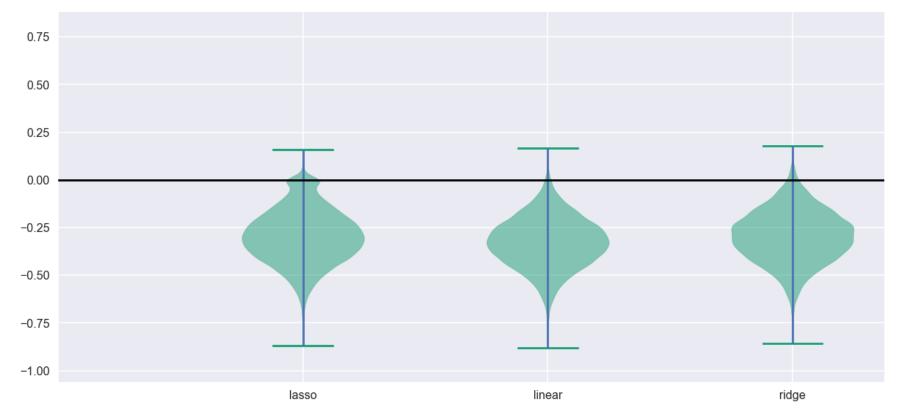


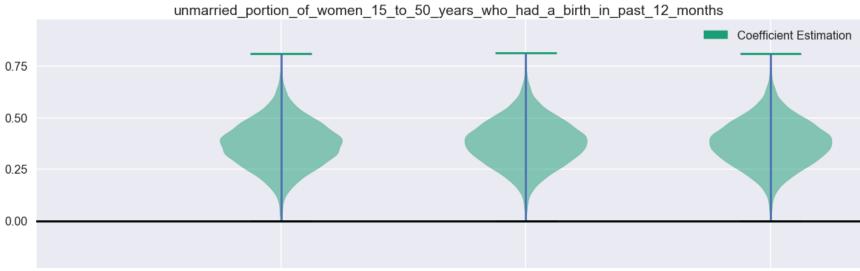
Confidence Results

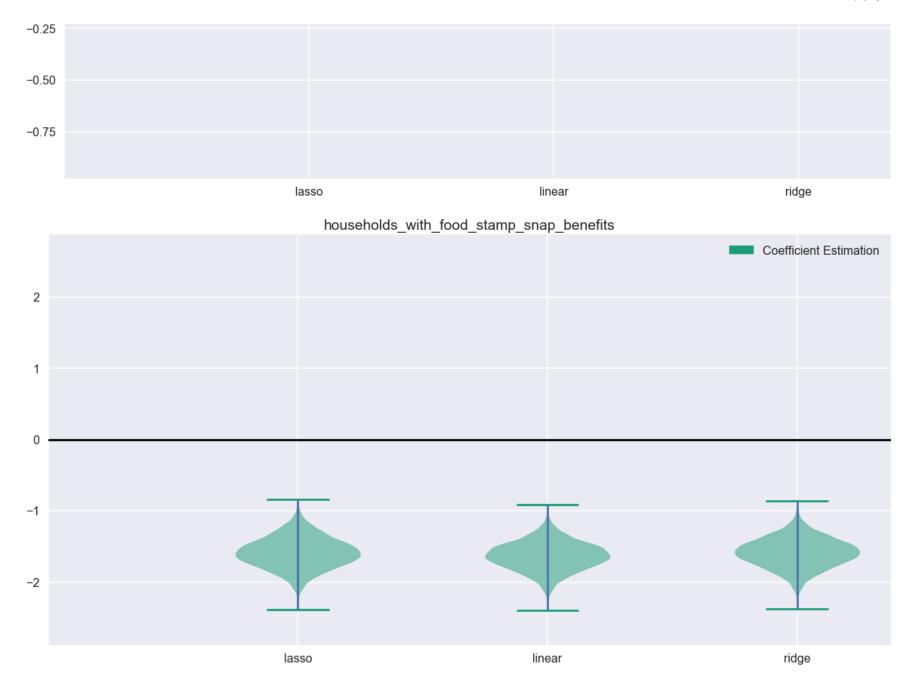












percentage married-couple family

