

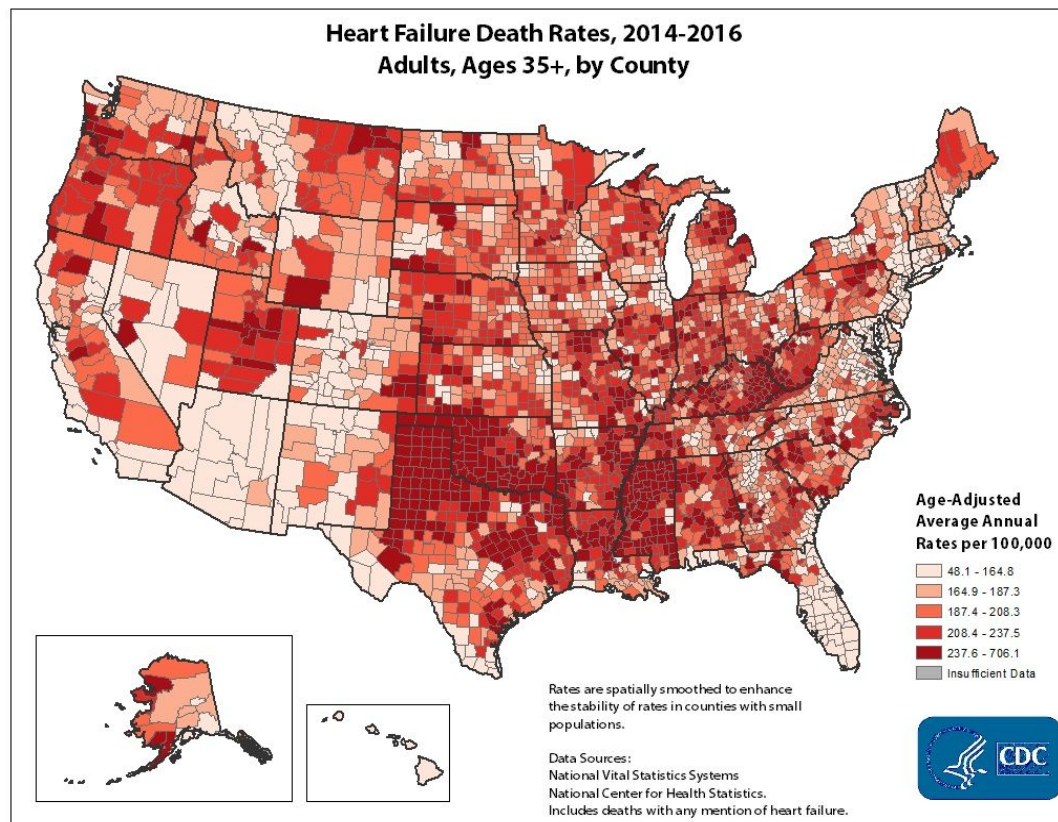
Heart Failure



Mark Perez

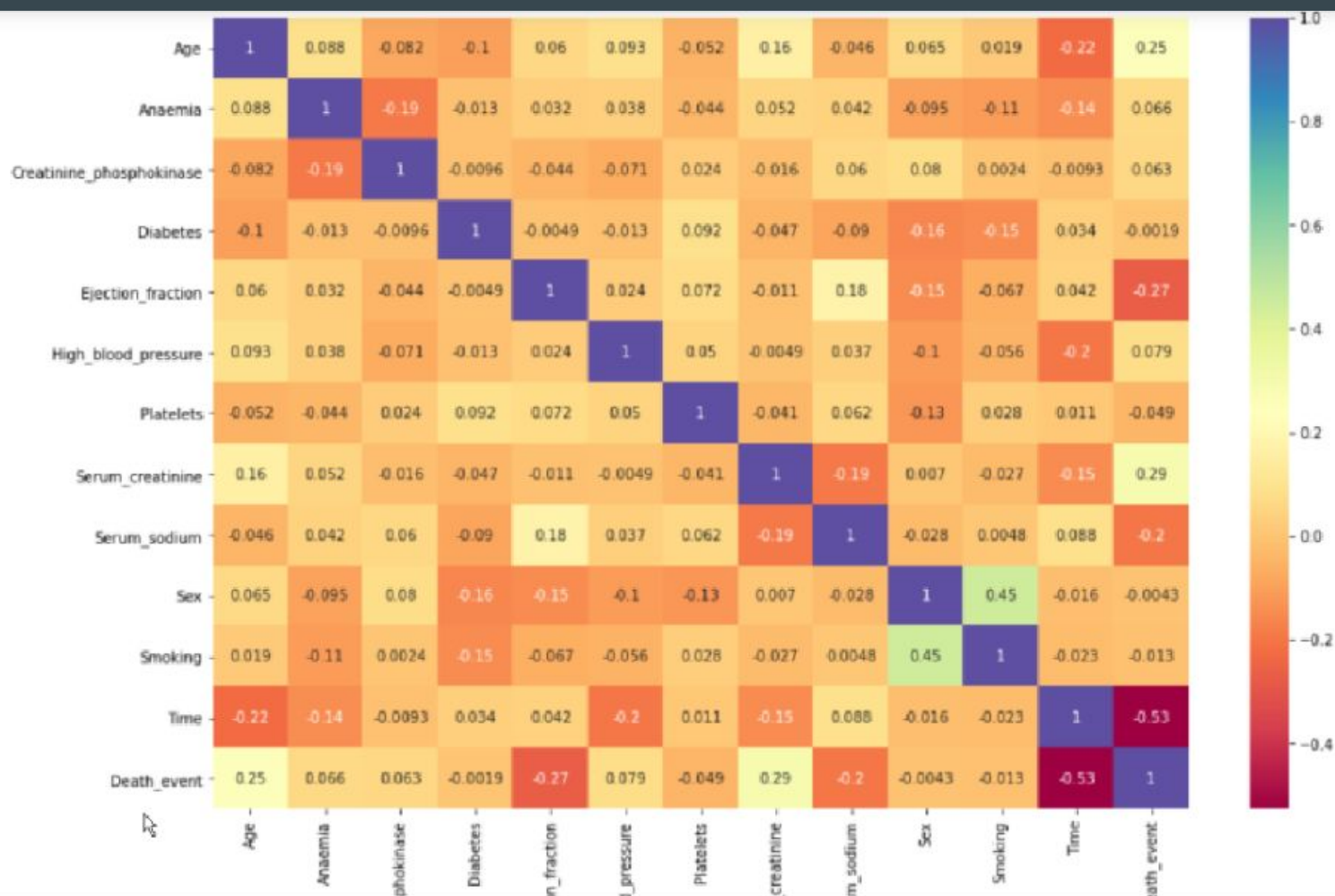
Heart Failure greatly affects us

About 6.2 million in United States.



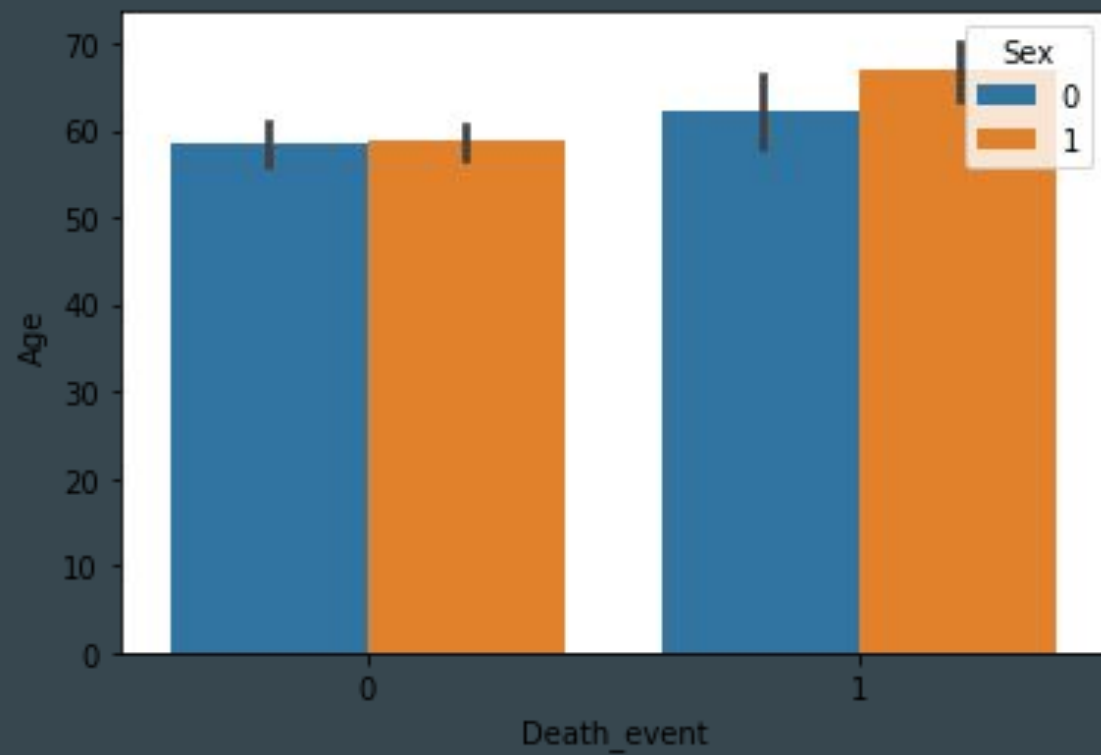
Can we predict what causes heart failure?

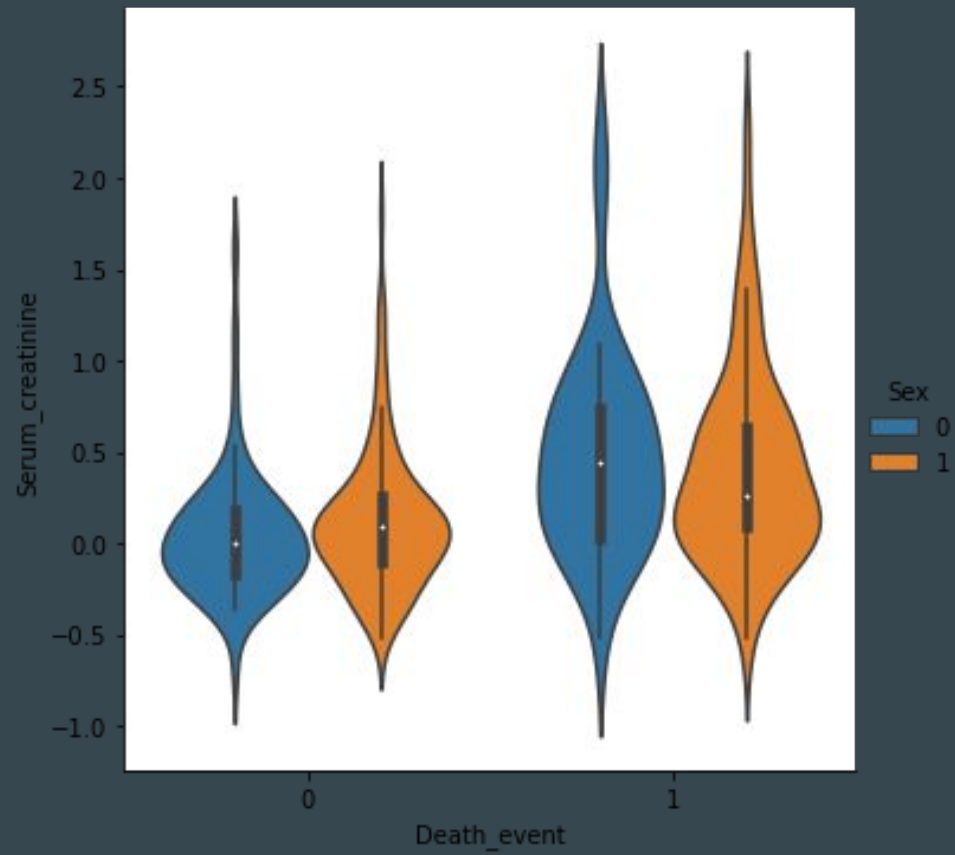
```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 299 entries, 0 to 298
Data columns (total 13 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   age                                   299 non-null    float64
1   anaemia                               299 non-null    int64
2   creatinine_phosphokinase              299 non-null    int64
3   diabetes                              299 non-null    int64
4   ejection_fraction                     299 non-null    int64
5   high_blood_pressure                   299 non-null    int64
6   platelets                             299 non-null    float64
7   serum_creatinine                       299 non-null    float64
8   serum_sodium                          299 non-null    int64
9   sex                                   299 non-null    int64
10  smoking                               299 non-null    int64
11  time                                  299 non-null    int64
12  DEATH_EVENT                           299 non-null    int64
dtypes: float64(3), int64(10)
memory usage: 30.5 KB
```

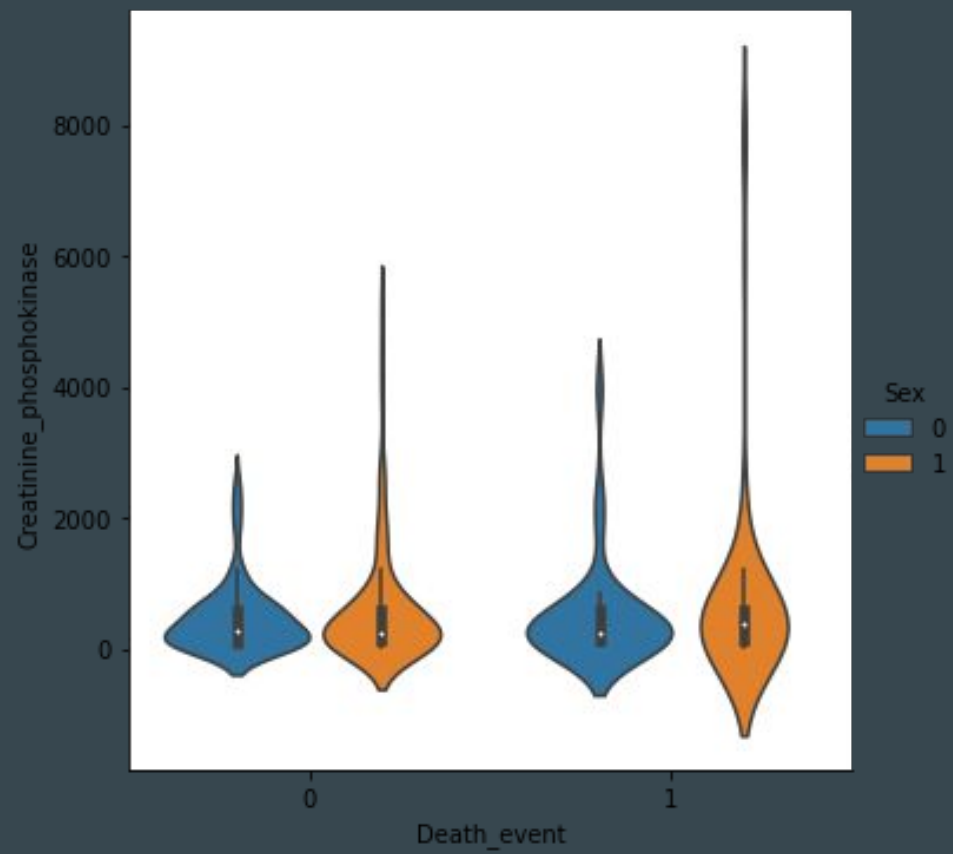


Interesting correlations

- Death_event - Serum_creatinine
- Death_event - Age
- Death_event - high_blood_pressure







Modeling

	Accuracy	Balanced Accuracy	ROC AUC	F1 Score	Time Taken
Model					
LinearSVC	0.79	0.76	0.76	0.78	0.01
LogisticRegression	0.79	0.76	0.76	0.78	0.01
NearestCentroid	0.77	0.75	0.75	0.76	0.01
XGBClassifier	0.77	0.74	0.74	0.76	1.16
CalibratedClassifierCV	0.78	0.74	0.74	0.76	0.03
RandomForestClassifier	0.77	0.74	0.74	0.76	0.11
ExtraTreesClassifier	0.77	0.72	0.72	0.75	0.09
LinearDiscriminantAnalysis	0.76	0.72	0.72	0.74	0.02
LGBMClassifier	0.74	0.72	0.72	0.74	0.03
RidgeClassifierCV	0.76	0.72	0.72	0.74	0.01
BaggingClassifier	0.74	0.72	0.72	0.74	0.02
RidgeClassifier	0.74	0.71	0.71	0.73	0.02
GaussianNB	0.73	0.70	0.70	0.72	0.01
SGDClassifier	0.72	0.70	0.70	0.72	0.01
SVC	0.73	0.70	0.70	0.72	0.01
AdaBoostClassifier	0.71	0.67	0.67	0.69	0.07
BernoulliNB	0.71	0.67	0.67	0.69	0.01
NuSVC	0.71	0.67	0.67	0.69	0.01
PassiveAggressiveClassifier	0.69	0.66	0.66	0.68	0.01
ExtraTreeClassifier	0.68	0.66	0.66	0.67	0.01
QuadraticDiscriminantAnalysis	0.69	0.65	0.65	0.67	0.01
DecisionTreeClassifier	0.68	0.65	0.65	0.67	0.01
Perceptron	0.67	0.63	0.63	0.65	0.01
KNeighborsClassifier	0.68	0.62	0.62	0.63	0.01
LabelPropagation	0.64	0.59	0.59	0.61	0.01
LabelSpreading	0.64	0.59	0.59	0.61	0.01
DummyClassifier	0.62	0.59	0.59	0.60	0.01

Hyperparameter tuning

- Selected certain parameters to tune and used RandomizedSearchCV.

How accurately can we predict?

Random Forest

```
In [31]: from sklearn.metrics import classification_report  
print(classification_report(y_pred,y_test))
```

	precision	recall	f1-score	support
0	0.94	0.72	0.82	69
1	0.49	0.86	0.62	21
accuracy			0.76	90
macro avg	0.71	0.79	0.72	90
weighted avg	0.84	0.76	0.77	90

Linear Regression

	precision	recall	f1-score	support
0	0.94	0.76	0.84	66
1	0.57	0.88	0.69	24
accuracy			0.79	90
macro avg	0.76	0.82	0.76	90
weighted avg	0.84	0.79	0.80	90

XGBClassifier

	precision	recall	f1-score	support
0	0.89	0.77	0.82	61
1	0.62	0.79	0.70	29
accuracy			0.78	90
macro avg	0.75	0.78	0.76	90
weighted avg	0.80	0.78	0.78	90

```
In [46]: for feature_list_index in sfm.get_support(indices=True):  
         print(feats_labels[feature_list_index])
```

Serum_creatinine
Time

```
In [48]: for feature_list_index in sfm.get_support(indices=True):  
         print(feats_labels[feature_list_index])
```

Time

Further Analysis

- Accumulate more data
- Explore other models
- Explore more hyperparameters