

# Feature Selection Tool

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## File:

feature\_selection.py

## main function:

```
SFE(df, estimator, param_grid, im_method = 'seq', sel_method = 'sfe',  
    vip_feat = 20, Forward = True, max_feat = 50, batch_feat = 1,  
    para_search_step = 1, random_state = None, n_jobs = 4,  
    cv = 5, randcv = True)
```

**PARAS1:** parameter of im\_method, get feature importance of each feature, support:

- 'fscore': F-score of features
- 'pcc': Person corr. Coefficient of features
- 'tree': optimal tree's feature importance(gini importance)
- 'lasso': weights of features for optimal lasso regression model
- 'elnet': weights of features for optimal elnet regression model
- 'seq': sequence of the features(default, namely feature importance is feature's sequence(order) )

**PARAS2:** parameter of sel\_method, method of get best feature numbers, support:

- 'sfe': Stepwise Feature Elimination, if Forward is True, then feature addition(add features one by one using recursive try algorithm)

- 'ofe': Ordered Feature Elimination, if Forward is True, then feature addition(add features one by one with the original order or feature importance order)

**PARAS3:** other parameters:

- vip\_feat = 20: very important features, will be fixed in the elimination or addition

- Forward = True: if forward is true, features will be added one by one instead of elimination

max\_feat = 50: max number of features  
batch\_feat = 1: batch size of features for addition or elimination  
para\_search\_step = 1: step of search best parameters during addition or elimination  
random\_state = None  
n\_jobs = 4: number of threads for parallelization  
cv = 5: cross validation's fold, default cross validated score is ROC-AUC for classification, R-squared for regression  
randcv = True: if True, then random search method will be used in the grid-search.

## Tests:

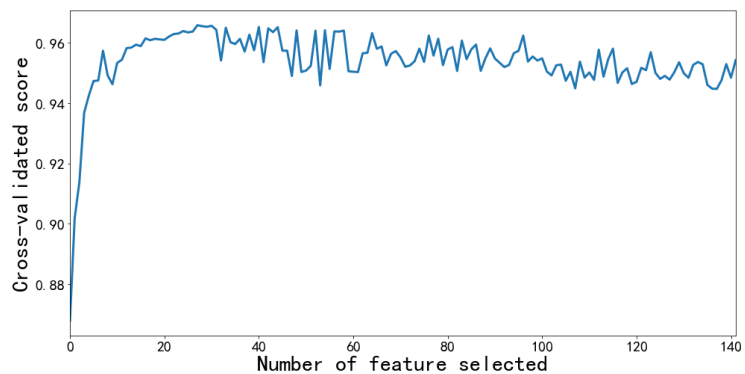
the tool has been tested on dataset of bairong(bairong\_train.csv)

### OFE test:

**estimator = SVM()**

**im\_method= 'tree'**

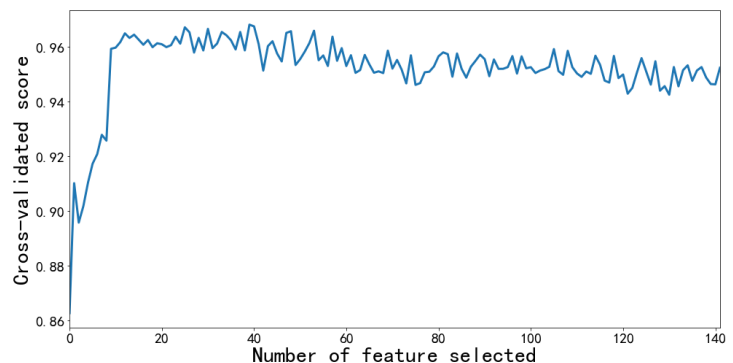
**sel\_method = 'ofe':**



**estimator = SVM()**

**im\_method= 'lasso'**

**sel\_method = 'ofe'**

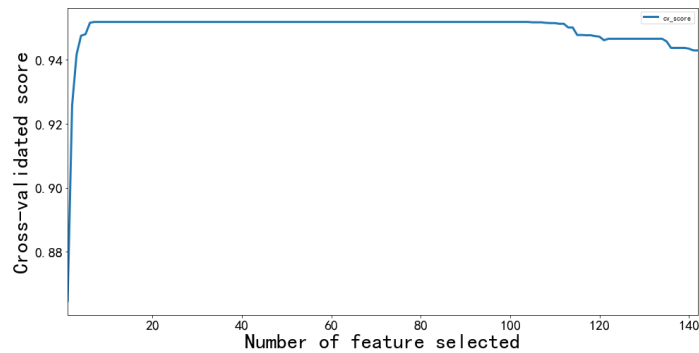


## SFE test:

```
estimator = DecisionTreeClassifier()
```

```
im_method= 'tree'
```

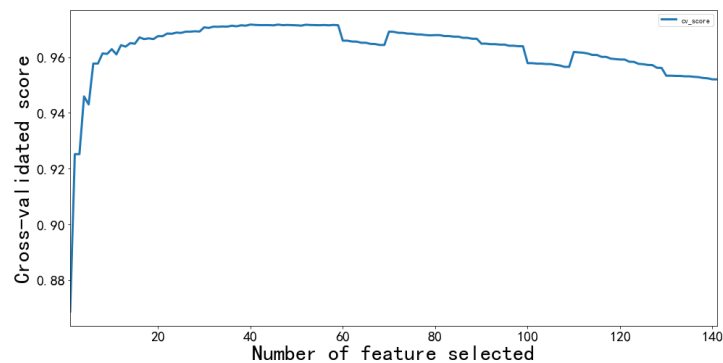
```
sel_method = 'sfe'
```



```
estimator = SVM()
```

```
im_method= 'tree'
```

```
sel_method = 'sfe'
```



## Conclusions:

- ◆ Both OFE and SFE can be used to select best number of features to avoid overfitting;
- ◆ SFE is a better method to select best number of features (Because SFE can raise cross validated score step by step)
- ◆ OFE is more faster than SFE