Embedded Methods

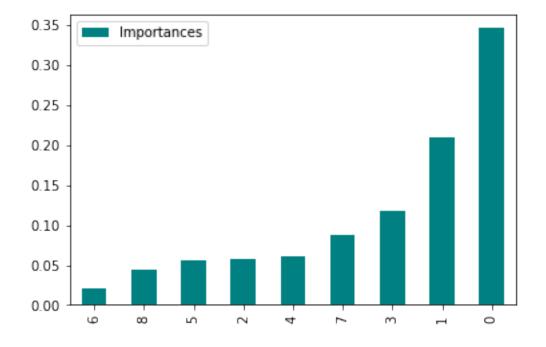
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[15]: # LASSO Regularization (L1)
      import numpy as np
      from sklearn.linear_model import LogisticRegression
      from sklearn.feature_selection import SelectFromModel
      import pandas as pd
      from sklearn.model selection import train test split
      from sklearn.preprocessing import StandardScaler
      dataset = pd.read_csv('dataset.csv')
      X= dataset.drop(columns='Result')
      Y= dataset['Result']
      # X.head()
      X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.2)
      # Scaling the data, as linear models benefits from feature scaling
      scaler = StandardScaler()
      scaler.fit(X_train.fillna(0))
      # set the regularization parameter c=1
      logistic = LogisticRegression(C=1, penalty='l1', solver='liblinear',_
      →random state=7)
      sel_ = SelectFromModel(logistic)
      sel_.fit(scaler.transform(X_train.fillna(0)), Y_train)
      sel_.get_support()
      selected_feat = X_train.columns[(sel_.get_support())]
      print('total features: {}'.format((X_train.shape[1])))
      print('selected features: {}'.format(len(selected_feat)))
      print('features with coefficients shrank to zero: {}'.format(
            np.sum(sel_.estimator_.coef_ == 0)))
     total features: 9
```

selected features: 9
features with coefficients shrank to zero: 0

```
[1]: #Random Forest Importance
     from sklearn.ensemble import RandomForestClassifier
     import pandas as pd
     from sklearn.model_selection import train_test_split
     dataset = pd.read_csv('dataset.csv')
     X= dataset.drop(columns='Result')
     Y= dataset['Result']
     # X.head()
     X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size=0.2)
     model = RandomForestClassifier(n_estimators=340)
     model.fit(X_train,Y_train)
     # get the importance of resulting features
     importances = model.feature_importances_
     # create a dataframe for vissualization
     final_df = pd.DataFrame({"Features": pd.DataFrame(X).columns, "Importances":
      →importances})
     final_df.set_index('Importances')
     # sort in ascending order for better vissualization
     final_df = final_df.sort_values('Importances')
     final_df.plot.bar(color='teal')
```

[1]: <matplotlib.axes._subplots.AxesSubplot at 0x241fc5c2688>



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