```
from google.colab import files
uploaded = files.upload()
```

选择文件 2 个文件

- **arr_big_data.csv**(application/vnd.ms-excel) - 141518235 bytes, last modified: 2020/4/30 - 100% done
- **dep_big_data.csv**(application/vnd.ms-excel) - 139548527 bytes, last modified: 2020/4/30 - 100% done

```
Saving arr_big_data.csv to arr_big_data.csv
Saving dep_big_data.csv to dep_big_data.csv
```

```
import pandas as pd
data = pd.read_csv('arr_big_data.csv')
pd.DataFrame.from_records(data)
# del data['Unnamed: 0']
# data
```

	presidents_day	easter	memorial_day	independence_day	labor_day	thanksgiving	win
0	0	0	0	0	0	0	
1	0	0	0	0	0	0	
2	0	0	0	0	0	0	
3	0	0	0	0	0	0	
4	0	0	0	0	0	0	
...	...	...	...	...	...	...	
177521	0	0	0	0	0	0	
177522	0	0	0	0	0	0	
177523	0	0	0	0	0	0	
177524	0	0	0	0	0	0	
177525	0	1	0	0	0	0	

177526 rows × 383 columns

Predictions for Arrival Delay

```
import pandas as pd
from sklearn.model_selection import train_test_split
import xgboost as xgb
from sklearn.metrics import mean_squared_error
import numpy as np

# data.dropna(axis=0, subset=['total_arr_delay'], inplace=True)

y = data.total_arr_delay
X = data.drop(['total_arr_delay'], axis=1).select_dtypes(exclude=['object'])

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=888)
```

Mean Absolute Error results

```
data_dmatrix = xgb.DMatrix(data=X,label=y)
# dtrain = xgb.DMatrix(X_train, label=y_train)
# dtest = xgb.DMatrix(X_test, label=y_test)

xg_reg = xgb.XGBRegressor(objective ='reg:linear', colsample_bytree = 0.3, learning_rate = 0.1,
                          max_depth = 5, alpha = 10, n_estimators = 10)

xg_reg.fit(X_train,y_train)

preds = xg_reg.predict(X_test)

rmse = np.sqrt(mean_squared_error(y_test, preds))
print("RMSE: %f" % (rmse))
```

```
RMSE: 23.745028
```

Using k-fold Cross Validation for model tuning

```
# params = {"objective":"reg:linear",'colsample_bytree': 0.3,'learning_rate': 0.1,
#                                 'max_depth': 5, 'alpha': 10}

# cv_results = xgb.cv(dtrain=data_dmatrix, params=params, nfold=3,
#                                     num_boost_round=50,early_stopping_rounds=10,metrics="rmse", as_panc

params = {"objective":'reg:squarederror','colsample_bytree': 0.3,'learning_rate': 0.35,
                                'max_depth': 7, 'alpha': 10, 'subsample': 0.7, 'gamma': 0.1}
cv_results_g1 = xgb.cv(dtrain=data_dmatrix, params=params, nfold=5,
                                    num_boost_round=200,early_stopping_rounds=10,metrics="rmse", as_pandas

cv_results_g1.head()
```

	train-rmse-mean	train-rmse-std	test-rmse-mean	test-rmse-std
0	29.512923	0.038624	29.551592	0.161476
1	23.782266	2.543996	23.860410	2.448174
2	22.744114	3.272639	22.890786	3.134779
3	21.819126	3.650699	22.018433	3.552576
4	21.699805	3.681405	21.923173	3.580048

Better RMSE

```
print((cv_results_g1["test-rmse-mean"]).tail(1))
```

```
199    13.856875
Name: test-rmse-mean, dtype: float64
```

```
xg_reg = xgb.train(params=params, dtrain=data_dmatrix, num_boost_round=10)
```

Visualize Boosting Trees and Feature Importance

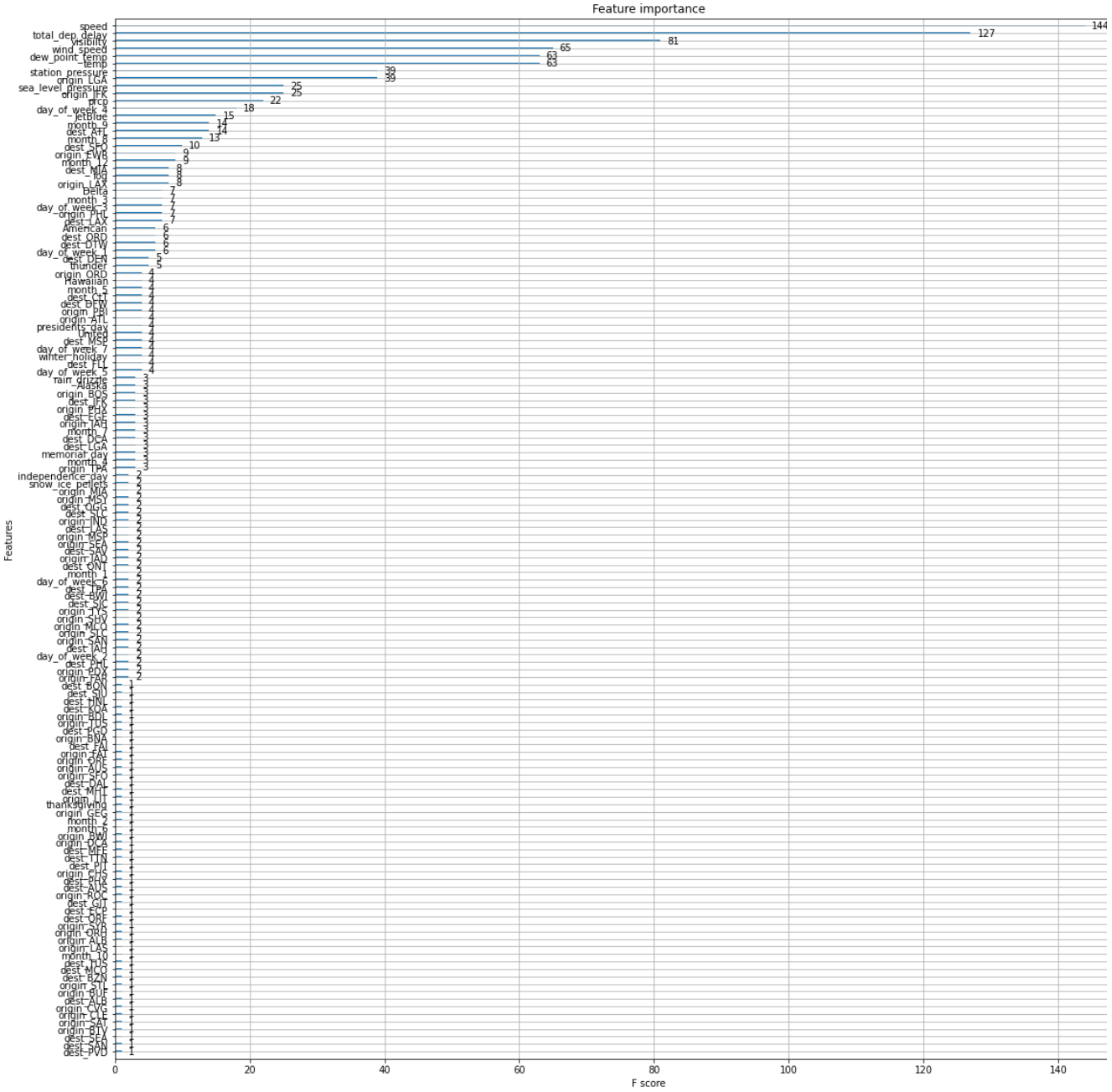
```
from  numpy  import  loadtxt
from  xgboost  import  XGBClassifier
from  xgboost  import  plot_tree
import  matplotlib.pyplot  as  plt

plt.rcParams["figure.figsize"]  =  20,20
plot_tree(xg_reg,  num_trees=0,  rankdir='LR')
#  xgb.plot_tree(xg_reg,num_trees=0)
#  xgb.plot_tree(xg_reg,  num_trees=0,  rankdir='LR')
#  plt.show()
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f1bb7c230b8>
```

Top Features

```
xgb.plot_importance(xg_reg)
plt.rcParams["figure.figsize"]  =  20,20
plt.show()
```


Feature importance
speed 144
total_dep_delay 127
visibility 81
wind_speed 65
dew_point_temp 63
temp 63
station_pressure 39
origin_LGA 39
sea_level_pressure 25
origin_JFK 25
hr0 22
day_of_week_4 18
JetBlue 15
month_9 14
dest_ATL 14
month_8 13
dest_SFO 10
origin_EWR 9
month_12 9
dest_MIA 8
fog 8
origin_LAX 8
Delta 7
month_3 7
day_of_week_3 7
origin_PHL 7
dest_LAX 7
American 6
dest_ORD 6
dest_DTW 6
day_of_week_1 6
dest_DEN 5
thunder 5
origin_ORD 4
Hawaiian 4
month_5 4
dest_CHT 4
dest_DFW 4
origin_PBI 4
origin_ATL 4
presidents_day 4
United 4
dest_MSP 4
day_of_week_7 4
winter_holiday 4
dest_FLL 4
day_of_week_5 4
rain_drizzle 3
Alaska 3
origin_BOS 3
dest_JFK 3
origin_PHX 3
dest_EYE 3
origin_IAD 3
month_7 3
dest_DEN 3
dest_LGA 3
memorial_day 3
month_4 3
origin_TPA 3
independence_day 2
snow_ice_pellets 2
origin_MIA 2
origin_MSY 2
dest_OGG 2
dest_STL 2
origin_IND 2
dest_LAS 2
origin_MSP 2
origin_SEA 2
dest_HNL 2
origin_IAD 2
dest_CNT 2
month_1 2
day_of_week_6 2
dest_TPA 2
dest_BWI 2
dest_SJC 2
origin_TYS 2
origin_SHV 2
origin_MCO 2
origin_STL 2
origin_SAN 2
dest_IAD 2
day_of_week_2 2
dest_PHL 2
origin_BOX 2
origin_FAR 2
dest_BQN 1
dest_SJU 1
dest_HNL 1
dest_ROA 1
origin_BTL 1
origin_TUS 1
dest_PCO 1
origin_BNA 1
dest_FAI 1
origin_FAT 1
origin_ORF 1
origin_ABQ 1
origin_SFO 1
dest_DAL 1
dest_MHT 1
origin_LIT 1
thanksgiving 1
origin_GEG 1
month_2 1
month_6 1
origin_BWI 1
origin_DCA 1
dest_MYF 1
dest_ITN 1
dest_PIT 1
origin_CHS 1
dest_RIC 1
dest_AUS 1
origin_ROC 1
dest_GIT 1
dest_ECP 1
dest_OBF 1
origin_SYR 1
origin_ORD 1
origin_ALB 1
origin_LAS 1
month_10 1
dest_TUS 1
dest_MCO 1
dest_BZN 1
origin_STL 1
origin_BUF 1
dest_ALB 1
origin_CVG 1
origin_CLE 1
origin_SAT 1
origin_BTV 1
dest_SEA 1
dest_SAN 1
dest_PVD 1
Features
0
20
40
60
80
100
120
140
F score

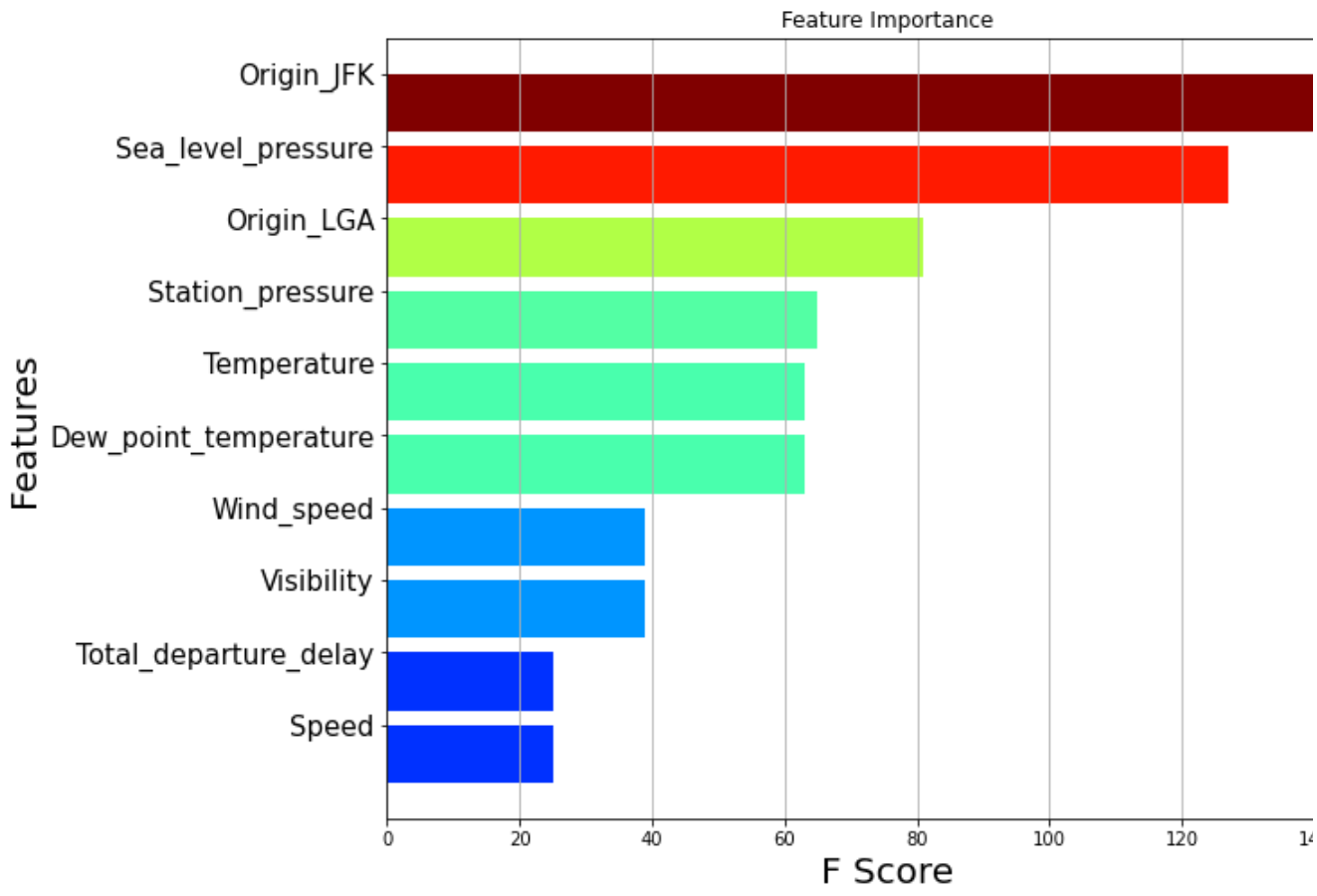
```
plt.figure(figsize=[10,8])

import matplotlib.pyplot as plt
from matplotlib import cm
import numpy as np

label = ['Speed','Total_departure_delay','Visibility','Wind_speed','Dew_point_temperature','Temperature','Sta
x = sorted([144,127,81,65,63,63,39,39,25,25])

idx = np.arange(len(x))
color = cm.jet(np.array(x)/max(x))
plt.barh(idx, x, color=color)
plt.yticks(idx+0.4,label,fontsize=15)
plt.grid(axis='x')
plt.xlabel('F Score',fontsize=20)
plt.ylabel('Features',fontsize=20)
plt.title('Feature Importance')

plt.show()
```



Predictions for Departure Delay

```
import pandas as pd
data = pd.read_csv('dep_big_data.csv')
pd.DataFrame.from_records(data)
```

	distance	presidents_day	easter	memorial_day	independence_day	labor_day	thanksg
0	479	0	0	0	0	0	
1	1222	0	0	0	0	0	
2	2381	0	0	0	0	0	
3	594	0	0	0	0	0	
4	1080	0	0	0	0	0	
...	...	...	...	...	...	...	
177521	594	0	0	0	0	0	
177522	695	0	0	0	0	0	
177523	1182	0	0	0	0	0	
177524	1846	0	0	0	0	0	
177525	546	0	1	0	0	0	

177526 rows × 382 columns

```
import pandas as pd
from sklearn.model_selection import train_test_split
import xgboost as xgb
from sklearn.metrics import mean_squared_error
import numpy as np

# data.dropna(axis=0, subset=['total_dep_delay'], inplace=True)

y = data.total_dep_delay
X = data.drop(['total_dep_delay'], axis=1).select_dtypes(exclude=['object'])


X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=123)
```

Mean Absolute Error results

```
data_dmatrix = xgb.DMatrix(data=X,label=y)

xg_reg = xgb.XGBRegressor(objective ='reg:linear', colsample_bytree = 0.3, learning_rate = 0.1,
                          max_depth = 5, alpha = 10, n_estimators = 10)


xg_reg.fit(X_train,y_train)

preds = xg_reg.predict(X_test)

rmse = np.sqrt(mean_squared_error(y_test, preds))
print("RMSE: %f" % (rmse))
```

```
RMSE: 27.676985
```

Using k-fold Cross Validation for model tuning

```
# params = {"objective":"reg:linear",'colsample_bytree': 0.3,'learning_rate': 0.1,
#                                 'max_depth': 5, 'alpha': 10}

# cv_results = xgb.cv(dtrain=data_dmatrix, params=params, nfold=3,
#                                         num_boost_round=50,early_stopping_rounds=10,metrics="rmse", as_panc

params = {"objective":'reg:squarederror','colsample_bytree': 0.3,'learning_rate': 0.1,
                                'max_depth': 7, 'alpha': 10, 'subsample': 0.7, 'gamma': 0.1}
cv_results_g1 = xgb.cv(dtrain=data_dmatrix, params=params, nfold=5,
                                      num_boost_round=200,early_stopping_rounds=10,metrics="rmse", as_pandas


cv_results_g1.head()
```

	train-rmse-mean	train-rmse-std	test-rmse-mean	test-rmse-std
0	33.669781	0.044570	33.679391	0.188015
1	32.377024	0.045918	32.402808	0.187011
2	31.299827	0.045561	31.335240	0.183580
3	30.384858	0.045082	30.436723	0.186172
4	29.630687	0.047559	29.693999	0.186524

Better RMSE

```
print((cv_results_g1["test-rmse-mean"]).tail(1))
```

```
199    25.592875
Name: test-rmse-mean, dtype: float64
```

```
xg_reg = xgb.train(params=params, dtrain=data_dmatrix, num_boost_round=10)
```

Visualize Boosting Trees and Feature Importance

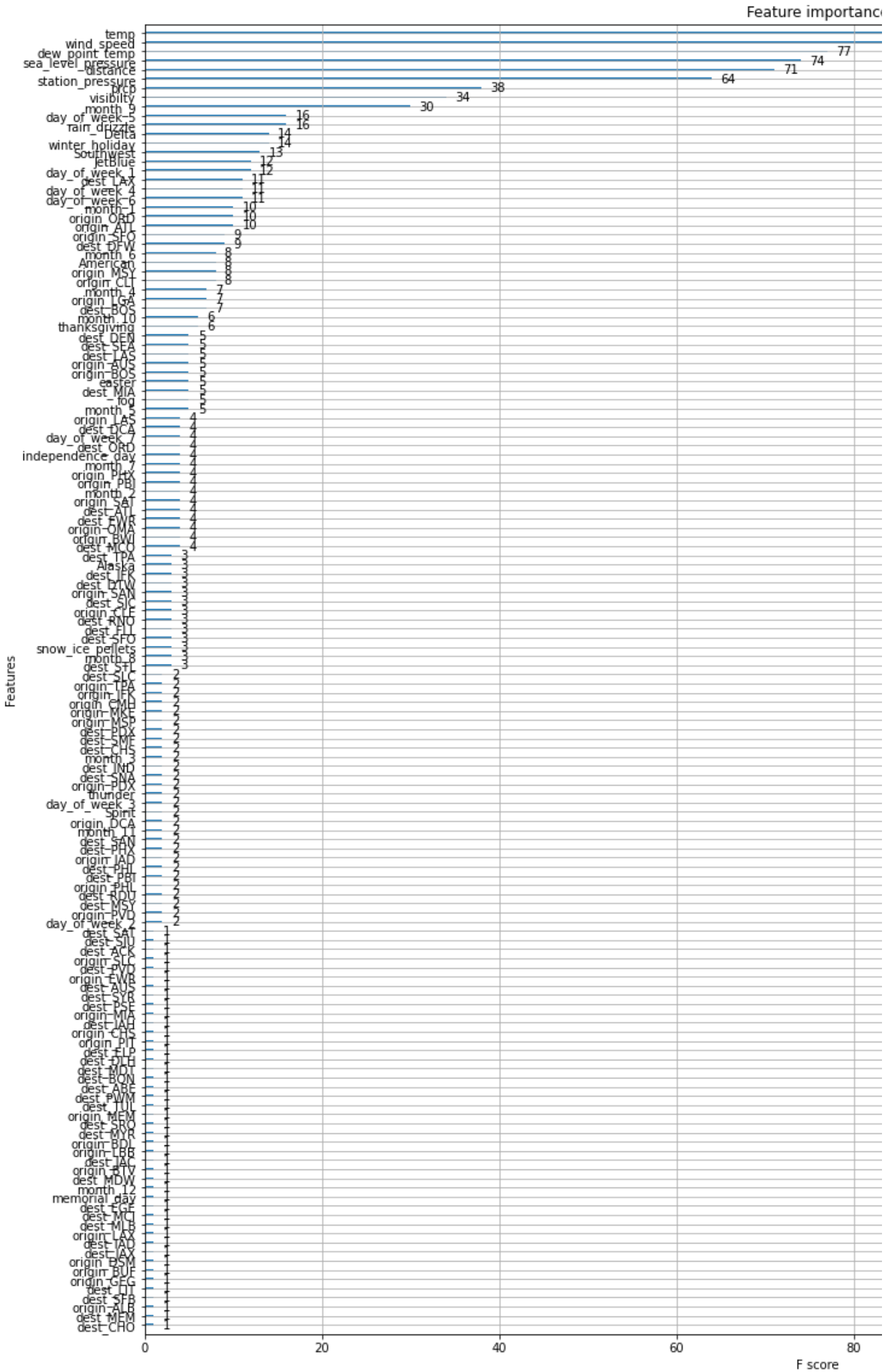
```
from numpy import loadtxt
from xgboost import XGBClassifier
from xgboost import plot_tree
import matplotlib.pyplot as plt

plt.rcParams["figure.figsize"] = 20,20
plot_tree(xg_reg, num_trees=0, rankdir='LR')
# xgb.plot_tree(xg_reg,num_trees=0)
# xgb.plot_tree(xg_reg, num_trees=0, rankdir='LR')
# plt.show()
```

```
<matplotlib.axes._subplots.AxesSubplot at 0x7f1bb5dec588>
```

Top Features

```
xgb.plot_importance(xg_reg)
plt.rcParams['figure.figsize'] = [5, 5]
plt.show()
```


Feature importance
Features
F score
0
20
40
60
80
temp
wind_speed
dew_point_temp
sea_level_pressure
distance
station_pressure
prcp
visibility
month_9
day_of_week_5
rain_drizzle
Delta
winter_holiday
Southwest
JetBlue
day_of_week_1
dest_LAX
day_of_week_4
day_of_week_6
month_1
origin_ORD
origin_ATL
origin_SFO
dest_DFW
month_6
American
origin_MSY
origin_CLT
month_4
origin_LGA
dest_BOS
month_10
thanksgiving
dest_DEN
dest_SEA
dest_LAS
origin_AUS
origin_BOS
easter
dest_MIA
fog
month_5
origin_LAS
dest_DCA
day_of_week_7
dest_ORD
independence_day
month_7
origin_PHX
origin_PBI
month_2
origin_SAT
dest_ATL
dest_EWR
origin_OMA
origin_BWI
dest_MCO
dest_TPA
Alaska
dest_JFK
dest_DTW
origin_SAN
dest_SJC
origin_CLE
dest_RNO
dest_FLL
dest_SFO
snow_ice_pellets
month_8
dest_STL
dest_SLC
origin_TPA
origin_IAH
origin_CMH
origin_MKE
origin_MSP
dest_PDX
dest_SMF
dest_CHS
month_3
dest_IND
dest_SNA
origin_PDX
thunder
day_of_week_3
Spirit
origin_DCA
month_11
dest_SAN
dest_PHX
origin_IAD
dest_PHL
dest_PBI
origin_PHL
dest_RDU
dest_MSY
origin_PVD
day_of_week_2
dest_SAT
dest_SJU
dest_ACK
origin_SLC
dest_PVD
origin_EWR
dest_AUS
dest_SYR
dest_PSE
origin_MIA
dest_IAH
origin_CHS
origin_PIT
dest_ELP
dest_DLH
dest_MDT
dest_BQN
dest_ABE
dest_PWM
dest_HOU
origin_MEM
dest_SRQ
dest_MYR
origin_BDL
origin_LBB
dest_IAD
origin_BUF
dest_MDW
month_12
memorial_day
dest_EGE
dest_MCI
dest_MLB
origin_LAX
dest_TAD
dest_JAX
origin_BSM
origin_BUF
origin_GEG
dest_CLT
dest_STT
origin_ALB
dest_MEM
dest_CHO