## In [1]:

```
import pandas as pd
import numpy as np
import xgboost as xgb
from sklearn.model_selection import train_test_split
import matplotlib.pyplot as plt

%matplotlib inline
plt.style.use('ggplot')
```

### In [2]:

```
df = pd.read_csv('Titanic.csv')
```

### In [3]:

```
df.head()
```

#### Out[3]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	Ci
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	1
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833	
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	1
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	С
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	1

# In [4]:

```
def proc(str1):
    res = str1.split(',')[1].split('.')[0].strip()
    if res in ['Mr', 'Miss', 'Mrs', 'Master']:
        return res
    else:
        return 'Others'

df['Title']=df['Name'].astype(str).apply(proc)
```

```
In [5]:
df['HasCabin']=df['Cabin'].apply(pd.isnull)
In [6]:
df['Sex'].value_counts()
Out[6]:
male
          577
female
          314
Name: Sex, dtype: int64
In [7]:
df['Title'].value counts()
Out[7]:
Mr
          517
          182
Miss
Mrs
          125
           40
Master
Others
           27
Name: Title, dtype: int64
In [8]:
df['Embarked'].value_counts()
Out[8]:
S
     644
С
     168
      77
Name: Embarked, dtype: int64
In [9]:
for t in ['Sex', 'Title', 'Embarked']:
    df = pd.concat([df, pd.get_dummies(df[t])], axis=1)
```

## In [10]:

```
df.head()
```

## Out[10]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare	
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500	
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	PC 17599	71.2833	
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250	
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000	
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500	

5 rows × 24 columns

## In [11]:

```
df.columns
```

### Out[11]:

### In [12]:

## In [13]:

```
df_train, df_test = train_test_split(df, stratify=df['Survived'], shuffle=True, test
```

#### In [14]:

```
data_trn = xgb.DMatrix(df_train[feats], label=df_train[target])
data_val = xgb.DMatrix(df_test[feats], label=df_test[target])
```

#### In [60]:

```
param = {
    'silent': 1,
    'eta': 0.08,
    'gamma': 0,
    'max_depth ': 2,
    'min_child_weight': 1,
    'subsample': 1,
    'lambda': 1,
    'alpha': 0,
    'bjective': 'binary:logistic',
    'eval_metric': 'logloss',
}
```

#### In [61]:

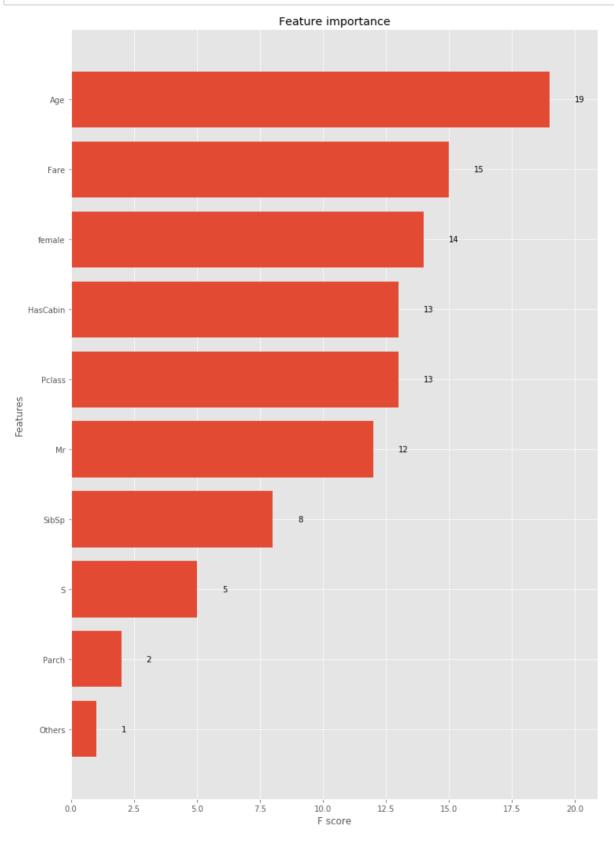
```
evals result = {}
bst = xgb.train(
    params=param, # Booster params
    dtrain=data trn, # Data to be trained
    num boost round=200, # Number of boosting iterations
    evals=[(data_trn, 'train'), (data_val, 'eval')], # List of items to be evaluated
    obj=None, # Customized objective function
    feval=None, # Customized evaluation function
   maximize=False, # Whether to maximize feval
    early stopping rounds=3, # Validation error needs to decrease at least every <ed
    evals result=evals result, # This dictionary stores the evaluation results of a
    verbose eval=2,
    learning rates=None, # List of learning rate for each boosting round
    xqb model=None,
   callbacks=None, # list of callback functions
)
```

[0] train-logloss:0.644084 eval-logloss:0.655657 Multiple eval metrics have been passed: 'eval-logloss' will be used fo r early stopping.

```
Will train until eval-logloss hasn't improved in 3 rounds.
[2]
        train-logloss:0.563202
                               eval-logloss:0.597299
[4]
        train-logloss:0.500736 eval-logloss:0.558651
        train-logloss:0.449635
                                eval-logloss:0.529011
[6]
        train-logloss:0.407089
                                eval-logloss:0.507065
[8]
        train-logloss:0.37451
                                eval-logloss:0.492354
[10]
        train-logloss:0.344772
                                eval-logloss:0.481165
[12]
        train-logloss:0.320689
                                eval-logloss:0.471064
[14]
        train-logloss:0.298508
[16]
                                eval-logloss:0.462935
        train-logloss:0.280579
                                eval-logloss:0.459645
[18]
[20]
        train-logloss:0.263187
                                eval-logloss:0.458922
        train-logloss:0.249087
                                eval-logloss:0.458764
[22]
Stopping. Best iteration:
[19]
        train-logloss:0.271463 eval-logloss:0.458588
```

## In [56]:

fig, ax = plt.subplots(figsize=(12,18))
xgb.plot\_importance(bst, max\_num\_features=50, height=0.8, ax=ax, importance\_type='we
plt.show()

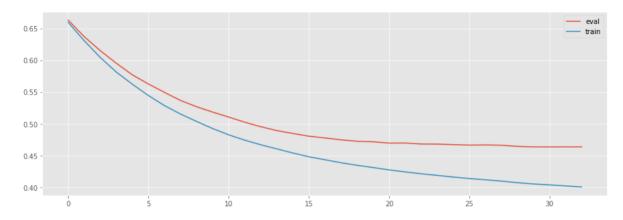


# In [57]:

pd.DataFrame({'eval': evals\_result['eval']['logloss'], 'train': evals\_result['train

# Out[57]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x1a1d4b4f28>



## In [ ]: