### **Get SparkContext**

### **Loading Artist-name data**

```
rawArtistData = sc.textFile("../Dropbox/pj_ss/audio/artist_data.txt")
In [3]:
In [4]: rawArtistData.take(10)
Out[4]: ['1134999\t06Crazy Life',
         '6821360\tPang Nakarin',
         '10113088\tTerfel, Bartoli- Mozart: Don',
         '10151459\tThe Flaming Sidebur',
         '6826647\tBodenstandig 3000',
         '10186265\tJota Quest e Ivete Sangalo',
         '6828986\tToto XX (1977',
         '10236364\tU.S Bombs -',
         '1135000\tartist formaly know as Mat',
         '10299728\tKassierer - Musik für beide Ohren'
In [5]: artistDataDF = rawArtistData\
            .map(lambda line: line.split("\t", 1))\
            .filter(lambda x: len(x) == 2)\
            .filter(lambda x: x[0].isdigit())\
            .map(lambda x: Row(artist=int(x[0]), name=x[1]))\
            .toDF()
In [6]: artistDataDF.take(2)
Out[6]: [Row(artist=1134999, name='06Crazy Life'),
         Row(artist=6821360, name='Pang Nakarin')]
```

# **Loading Artist alias**

```
In [7]: rawArtistAlias = sc.textFile('../Dropbox/pj_ss/audio/artist_alias.txt')
```

```
In [8]: rawArtistAlias.take(10)
 Out[8]: ['1092764\t1000311',
           '1095122\t1000557',
           '6708070\t1007267',
           '10088054\t1042317',
           '1195917\t1042317',
           '1112006\t1000557',
           '1187350\t1294511'
           '1116694\t1327092',
           '6793225\t1042317',
           '1079959\t1000557']
 In [9]: artistAliasDF = rawArtistAlias\
              .map(lambda line: line.split("\t"))\
              .filter(lambda x: len(x) == 2)\
              .filter(lambda x: x[0].isdigit())\
              .filter(lambda x: x[1].isdigit())\
              .map(lambda x: Row(artist=int(x[0]), alias=int(x[1])))\
              .toDF()
In [10]: artistAliasDF.take(2)
Out[10]: [Row(alias=1000311, artist=1092764), Row(alias=1000557, artist=109512
         2)]
```

## **Loading User-Artist data**

```
In [16]: print(userArtistDF.select('user').rdd.max())
    print(userArtistDF.select('user').rdd.min())
    print(userArtistDF.select('artist').rdd.max())
    print(userArtistDF.select('artist').rdd.min())

Row(user=2443548)
    Row(user=90)
    Row(artist=10794401)
    Row(artist=1)
```

## **Resolve Aliasing**

```
In [17]: trainData = userArtistDF.join(artistAliasDF, "artist", "left_outer")\
              .rdd\
              .map(lambda x: (x['user'], x['artist'], x['count']) if x['alias']==
         None \
                   else (x['user'], x['alias'], x['count']))\
              .map(lambda x: (int(x[0]), int(x[1]), int(x[2])))
In [18]: trainData.take(10)
Out[18]: [(1000159, 26, 1),
           (1000320, 26, 1),
          (1000385, 26, 1),
           (1000404, 26, 3),
           (1000405, 26, 9),
          (1000415, 26, 1),
           (1000518, 26, 1),
          (1000573, 26, 63),
          (1000596, 26, 3),
          (1000626, 26, 5)]
In [19]: trainData.cache()
Out[19]: PythonRDD[69] at RDD at PythonRDD.scala:49
```

### **Build Model**

```
In [20]: from pyspark.mllib.recommendation import ALS, MatrixFactorizationModel,
    Rating

rank = 10
    iterations = 5
    lbda = 0.01
    seed = 42
```

```
In [22]: model
```

Out[22]: <pyspark.mllib.recommendation.MatrixFactorizationModel at 0x1064c3eb8>

### Model의 사용

```
In [23]: model.userFeatures().take(1)
Out[23]: [(7400,
           array('d', [-0.5253109335899353, -3.717756509780884, 0.55670106410980
         22, -0.6934528350830078, 1.8062853813171387, -0.573974609375, 2.7792887
         687683105, \ 0.9534238576889038, \ -2.383761405944824, \ -0.750530898571014
         4]))]
In [24]: model.productFeatures().take(1)
Out[24]: [(100,
           array('d', [-0.08007114380598068, -0.13515709340572357, -0.0206525325
         7751465, -0.008643229492008686, 0.03264828398823738, 0.0178102459758520
         10340531915426254]))]
In [25]: productId = 1
         model.recommendUsers(productId, 5)
Out[25]: [Rating(user=1037240, product=1, rating=1.3606984320279356),
         Rating(user=1054417, product=1, rating=1.336792097354692),
         Rating(user=2167160, product=1, rating=1.3278224025483398),
         Rating(user=1005353, product=1, rating=1.3089582300854892),
         Rating(user=1001440, product=1, rating=1.3073439566057818)]
In [26]: userId = 2010581
         model.recommendProducts(userId, 5)
Out[26]: [Rating(user=2010581, product=1077309, rating=1.2734482951652937),
         Rating(user=2010581, product=1010524, rating=1.244074256266154),
         Rating(user=2010581, product=1000602, rating=1.2122672161252315),
         Rating(user=2010581, product=1000175, rating=1.2115658940753997),
         Rating(user=2010581, product=1010267, rating=1.198888835210903)]
In [27]: model.predict(userId, productId)
Out[27]: 0.5197372067168629
```

++	+
name	artist
++	+
Chevelle	1000175
Zebrahead	1010267
Billy Talent	1077309
MC Chris	1010524
Finger Eleven	1000602
++	+

#### **Calculate MSE**

### Save and Load for later use

```
In [35]: model.save(sc, '../pyspark/recsysmodel')
In [36]: sameModel = MatrixFactorizationModel.load(sc, '../pyspark/recsysmodel')
```