Part 1:



Part 2:

- 1. fixed numIterations = 20
 - a. rank = 5

```
In [9]: 1 rank = 5
2 numIterations = 20
3 model = ALS.train(train_ratings, rank, numIterations)

In [10]: 1 testdata = train_ratings.map(lambda p: (p[0], p[1]))
predictions = model.predictAll(testdata).map(lambda r: ((r[0], r[1]), r[2]))
3 ratesAndPreds = train_ratings.map(lambda r: ((r[0], r[1]), r[2])).join(predictions)
4 MSE = ratesAndPreds.map(lambda r: (r[1][0] - r[1][1])**2).mean()
print("Mean Squared Error = " + str(MSE))
Mean Squared Error = 0.6187260909958737
```

b. rank = 10

c. rank = 20

```
In [13]: 1    rank = 20
2    numIterations = 20
3    model = ALS.train(train_ratings, rank, numIterations)

In [14]: 1    testdata = train_ratings.map(lambda p: (p[0], p[1]))
2         predictions = model.predictAll(testdata).map(lambda r: ((r[0], r[1]), r[2]))
3         ratesAndPreds = train_ratings.map(lambda r: ((r[0], r[1]), r[2])).join(predictions)
4         MSE = ratesAndPreds.map(lambda r: (r[1][0] - r[1][1])**2).mean()
5         print("Mean Squared Error = " + str(MSE))
Mean Squared Error = 0.29089591114428714
```

d. rank = 30

Mean Squared Error = 0.18314214236371854

As it shows, as rank increased, the MSE decreased.

2. fixed rank = 20

a. numIterations = 2

```
In [3]:
                 2 numIterations = 2
                3 model = ALS.train(train_ratings, rank, numIterations)
 In [4]: 1 testdata = train_ratings.map(lambda p: (p[0], p[1]))
                predictions = model.predictAll(testdata).map(lambda r: ((r[0], r[1]), r[2]))
ratesAndPreds = train_ratings.map(lambda r: ((r[0], r[1]), r[2])).join(predictions)
                4 MSE = ratesAndPreds.map(lambda r: (r[1][0] - r[1][1])**2).mean()
5 print("Mean Squared Error = " + str(MSE))
              Mean Squared Error = 0.5235247099214314
               b. numIterations = 5
In [7]: 1 rank = 20
                numIterations = 5
model = ALS.train(train_ratings, rank, numIterations)
In [8]: 1 testdata = train_ratings.map(lambda p: (p[0], p[1]))
2 predictions = model.predictAll(testdata).map(lambda r: ((r[0], r[1]), r[2]))
3 ratesAndPreds = train_ratings.map(lambda r: ((r[0], r[1]), r[2])).join(predictions)
4 MSE = ratesAndPreds.map(lambda r: (r[1][0] - r[1][1])**2).mean()
5 print("Mean Squared Error = " + str(MSE))
              Mean Squared Error = 0.337948246332415
               c. numlterations = 10
                 2 numIterations = 10
                 3 model = ALS.train(train_ratings, rank, numIterations)
In [10]: 1 testdata = train_ratings.map(lambda p: (p[0], p[1]))
                predictions = model.predictAll(testdata).map(lambda r: ((r[0], r[1]), r[2]))
ratesAndPreds = train_ratings.map(lambda r: ((r[0], r[1]), r[2])).join(predictions)

MSE = ratesAndPreds.map(lambda r: (r[1][0] - r[1][1])**2).mean()
print("Mean Squared Error = " + str(MSE))
```

d. numIterations = 20

Mean Squared Error = 0.30763536590666024

e. numIterations = 30

because numlteration is too large, the function cannot run properly.

As we see, as numlteration increased, the MSE decreased.

3. With fixed rank = 20, numIterations = 30

Size of data

Mean Squared Error

2000

0.291120758082147923

5000	0.291259682781938134
10000	0.29208892894787620
20000	0.291487303979932457
50000	0.291021837001804823
100000	0.291294923919096340

As data shows, the size of data won't affect MSE.