# Project documentation

Business Intelligence Project Lab 3

docker run --hostname=quickstart.cloudera --privileged=true -t -i -p 8888 cloudera/quickstart:latest /usr/bin/docker-quickstart

Run gradle runAllMapReduceJobs in the project folder

### map reduce 1

#### Implementation:

- gradle as build system (gradle)
- mapper to parse the file and map to (key: product, value: 1)
- reducer which produces (key: product, value: avg(product)) which calculates the arithmetic mean per product

#### Scalability:

- $\bullet\,$  expected speedup when run on multiple machines
  - o a bit less than linear due to cooridnation overhead

Result: first 15 scores per category

#### Amazon Instant Video:

```
B000GFDAUG 5.0
B000GIOPK2 4.351351351352
B000GIPKWY 3.857142857142857
B000GJUQ7M 3.66666666666665
B000GK0NBK 3.5
B000GK51HG 5.0
B000GK6NFK 4.389830508474576
B000GK7DPY 3.66666666666665
B000GOTJGG 2.333333333333333
B000GOV10S 5.0
B000GOW7RE 5.0
B000GOW9B8 3.0
B000GOYLNC 5.0
B000GP0T82 3.0
B000GP38JE
          5.0
```

#### Sports and Outdoors:

```
      0000031852
      3.857142857142857

      0000031895
      3.666666666666665

      0000031909
      2.6666666666666665

      0000032034
      3.66666666666666665

      0000032050
      4.75

      0000032069
      4.0

      0188477284
      2.0

      0531904822
      5.0
```

#### Video Games:

```
0078764343
          4.66666666666667
043933702x
            4.0
0439339960
0439339987 5.0
0439342260 4.0
0439374391 5.0
0439394422
043940133x 3.2857142857142856
0439573947 4.66666666666667
0439591295 5.0
0439591368
          3.75
0439591538 2.5
0439671418 5.0
0439715571
            4.0
0439773660
            2.0
```

## map reduce 2 sentiment analysis

#### Implementation:

- gradle as build system
- mapper to parse the file and map to (key: product, value1: numberPosWords, value2: numberNegWords)
- reducer which produces (key: product, value: sentiment(product, #posWords, #negWords)) which calculates the sentiment score per product

#### Questions:

- How many invocations of map reduce are there:
  - o Amazon Instant Video: 37126 and 1685
  - o Sports and Outdoors: 296337 and 18357
  - o Video Games: 231780 and 10672
- · where is most of the runtime spent
  - o parsing the json, disk IO
- what is the expected speedup when run on multiple machines
  - o less than linear due to shuffle overhead (network is slow)

Result: first 10 products per category

#### Sports and Outdoors:

```
1881509818 SentimentWritable [positive=17, negative=10, sentiment=0.25925925925925925924]
2094869245 SentimentWritable [positive=39, negative=8, sentiment=0.6595744680851063]
7245456259 SentimentWritable [positive=28, negative=7, sentiment=0.6]
7245456313 SentimentWritable [positive=497, negative=168, sentiment=0.49473684210526314]
B000002NUS SentimentWritable [positive=53, negative=28, sentiment=0.30864197530864196]
```

```
RUUUUUELZ5
              Sentimentwritable [positive=20, negative=22, sentiment=0.083333333333333333]
B00000TURU
              SentimentWritable [positive=32, negative=5, sentiment=0.7297297297297297]
              SentimentWritable [positive=38, negative=4, sentiment=0.8095238095238095]
B00000TUX5
B00000J6J0
              SentimentWritable [positive=148, negative=73, sentiment=0.3393665158371041]
B0000224UE
              SentimentWritable [positive=100, negative=34, sentiment=0.4925373134328358]
Amazon Instant Video:
B000H00VBQ
              SentimentWritable [positive=31, negative=39, sentiment=-0.11428571428571428]
в000н0х790
              SentimentWritable [positive=12, negative=3, sentiment=0.6]
B000H29TXU
              SentimentWritable [positive=9, negative=3, sentiment=0.5]
              SentimentWritable [positive=32, negative=12, sentiment=0.4545454545454545453]
B000H2DMME
B000H4YNM0
              SentimentWritable [positive=100, negative=88, sentiment=0.06382978723404255]
              SentimentWritable [positive=212, negative=155, sentiment=0.1553133514986376]
B000HAB4NK
B000HKWE30
              SentimentWritable [positive=11, negative=2, sentiment=0.6923076923076923]
B000HZEHL6
              SentimentWritable [positive=104, negative=82, sentiment=0.11827956989247312]
B000I5PVD8
              SentimentWritable [positive=27, negative=9, sentiment=0.5]
B000I5Q0ZG
              SentimentWritable [positive=13, negative=4, sentiment=0.5294117647058824]
Video games:
0700099867
              SentimentWritable [positive=158, negative=110, sentiment=0.1791044776119403]
6050036071
              SentimentWritable [positive=33, negative=15, sentiment=0.375]
7100027950
              SentimentWritable [positive=50, negative=61, sentiment=-0.0990990990990991]
              SentimentWritable [positive=18, negative=13, sentiment=0.16129032258064516]
7293000936
8176503290
              SentimentWritable [positive=58, negative=31, sentiment=0.30337078651685395]
907843905X
              SentimentWritable [positive=33, negative=20, sentiment=0.24528301886792453]
9625990674
              SentimentWritable [positive=76, negative=21, sentiment=0.5670103092783505]
9861019731
              SentimentWritable [positive=36, negative=5, sentiment=0.7560975609756098]
9882155456
              SentimentWritable [positive=91, negative=52, sentiment=0.2727272727272727]
B000003SQQ
              SentimentWritable [positive=50, negative=27, sentiment=0.2987012987012987]
```

### hive

#### RESULTS FOR ALL QUERIES

How many movies are there in total in the dataset?

```
32204001
1 Job
```

Map Reduce = sum

How many movies in the dataset belong to the "Film-Noir" genre?

```
233
1 job
Where, Sum
```

Which are the 10 most frequently assigned tags (by users, i.e., from the tags table)?

```
    sci-fi 3384
    based on a book 3281
    atmospheric 2917
    comedy 2779
```

```
    action 2657
    surreal 2427
    BD-R 2334
    twist ending 2323
    funny 2072
    dystopia 1991
```

Job 1: Group and calculate SUM Job 2: Sort and Limit

Which 10 movies were the most controversial in 2015 (i.e., had the highest variance in ratings between 2015/01/01 and 2015/12/31)?

```
45533
  6051 4
2
  2298 4
  101971 4
4
5
  128425
  126927 4
6
7
  3905 4
  72360
8
         4
9
  128173 4
10 128169 4
```

2 jobs

Job 1: Group, Variance (AVG) Job 2: Variance, Order

Which movies (titles) are the 10 most frequently tagged and how often have they been tagged?

```
Pulp Fiction (1994)
                       1994
2
    Fight Club (1999)
                      1779
3
   Inception (2010) 1552
    "Matrix 1430
    "Shawshank Redemption 1339
5
6
    Eternal Sunshine of the Spotless Mind (2004)
                                              1240
7
   Donnie Darko (2001) 1177
8
   Memento (2000) 1168
9
    "Silence of the Lambs 1100
10
   Avatar (2009) 995
```

2 jobs

Job 1: Group and calculate SUM Job 2: Join, Sort and Limit

Which 15 movies (titles) have been most frequently tagged with the label "mars"?

```
1
    Mars Attacks! (1996)
2
    "War of the Worlds 25
3
   Total Recall (2012)
4
   Capricorn One (1978)
5
   Total Recall (1990)
6
   Martian Child (2007) 6
    It Came from Outer Space (1953)
8
    Mission to Mars (2000) 4
9
    "Day the Earth Stood Still
   RocketMan (a.k.a. Rocket Man) (1997)
10
     "6th Day
11
12 Red Planet (2000)
13 Destination Moon (1950)
14
     Impostor (2002)
     D--- /000E1
```

```
10 DOOM (CUUD) 1
```

2 jobs

Job 1: Where, group and calculate SUM Job 2: Join, Sort and Limit

## Which are the 10 best-rated movies (on average; list titles) with more than 1000 ratings?

```
1
    Pulp Fiction (1994) 4.2625 67310
2
    Forrest Gump (1994) 4.1235 66172
3
    "Shawshank Redemption 4.539 63366
  "Silence of the Lambs 4.2695 63299
  Jurassic Park (1993) 3.7567 59715
5
   Star Wars: Episode IV - A New Hope (1977)
7
   Braveheart (1995)
                     4.1327 53769
   Terminator 2: Judgment Day (1991) 4.0214 52244
8
9
    "Matrix 4.3169 51334
   Schindler's List (1993) 4.4015 50054
1.0
```

2 jobs

Job 1: group and calculate SUM Job 2: Having, Join, Sort and Limit

## Which are the highest-rated "Film-Noir" movies with more than 10 ratings (average rating; movies with genre "Film-Noir", max. 10)?

```
1
    L.A. Confidential (1997) 4.1845
2
   Sin City (2005) 4.0856 15481
  Chinatown (1974) 4.2995 15310
  Dark City (1998) 3.9311 11759
4
5
   Mulholland Drive (2001) 3.986 9307
   Sunset Blvd. (a.k.a. Sunset Boulevard) (1950) 4.3537 6525
6
   Miller's Crossing (1990) 4.0957 6060
8
  Strangers on a Train (1951) 4.2631 5154
                              5053
   Blood Simple (1984) 4.1587
9
10
   Notorious (1946) 4.2847 4932
```

2 jobs

Job 1: where, group and calculate SUM Job 2: Having, Join, Sort and Limit

## What are the 15 most relevant genome tags for the movie "Toy Story (1995)" (movieId=1)?

```
1
          0.99924999999999997
    tovs
    computer animation     0.9984999999999999
2
3
    pixar animation 0.996
4
    kids and family 0.9907500000000002
5
    animation 0.9857499999999999
    kids 0.9792499999999995
7
    pixar 0.96675
    children 0.96425000000000005
    cartoon 0.9564999999999991
9
     imdb top 250     0.9419999999999999
10
11 animated 0.9332499999999991
12 childhood 0.92625000000000002
      great movie 0.9207499999999999
13
14
     disney animated feature 0.91375000000000006
15
     friendship 0.91175000000000006
1 job
```

job

Where, Join, Order, Limit (kein Group)

Which are the 10 most relevant movies for Vienna (i.e., with the highest genome tag relevance rating for the tag "vienna")?

```
"Third Man 0.98750000000000004
1
   Johnny Guitar (1954) 0.9664999999999991
2
  Before Sunrise (1995) 0.9612499999999994
  Before Sunset (2004) 0.9570000000000007
    Before Midnight (2013) 0.9110000000000003
5
    "Night Porter 0.8934999999999996
6
7
    "Illusionist 0.84525000000000006
   Amadeus (1984) 0.8414999999999991
8
    "Foreign Affair 0.7322499999999999
   Love in the Afternoon (1957) 0.66900000000000004
10
```

1 job

Where, Join, Order, Limit (kein Group)

# DESCRIPTION OF YOUR UNDERSTANDING OF WHAT HAPPENS BEHIND THE SCENES

including a discuss on how many MR jobs your queries are translated into and why. Finally, also comment on what scale-up you would expect when running your queries on a real cluster in paralle

#### Scalability:

- expected speedup when run on multiple machines
  - o a bit less than linear due to cooridnation overhead