

# Implement Distributed Alternating Least Squares Algorithm for Matrix Completion

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## **Netflix Problem**

- V: m\*n matrix
- complete the matrix

|         | Avatar | The Matrix | Up   |
|---------|--------|------------|------|
| Alice   | ( ?    | 4          | 2    |
| Bob     | 3      | 2          | ? ]. |
| Charlie |        | ?          | 3 /  |

- W: m\*r (row-factor matrix)
- H: r\*n (column-factor matrix)
- W\*H approx V
- Loss function (V<sub>ij</sub> WH<sub>ij</sub>)<sup>2</sup>



## **Motivation**

Large applications involve matrices with

- millions of rows x columns;
- billions of entries

To achieve high-performance

- parallel & distributed factorisation
- keep the loss to minimum



## Algorithm

#### **Sequential Computation**

- Initial point  $W_0$  and  $H_0$
- ALS solved for every row & column

Compute 
$$\boldsymbol{W}_{n+1}$$
:  $(\forall i) \ \underline{\boldsymbol{W}_{i*}} \boldsymbol{H}_n^{(i)} = \boldsymbol{V}_{i*},$   
Compute  $\boldsymbol{H}_{n+1}$ :  $(\forall j) \ \boldsymbol{W}_{n+1}^{(j)} \underline{\boldsymbol{H}_{*j}} = \boldsymbol{V}_{*j},$ 

#### **Parallel Computation**

Parallelise computation for rows and columns respectively



# Algorithm

#### **Distributed Computation**

- Partition (block) the matrix with  $m_b*n_b$  matrices
- every node updates a matrix block

#### Why Spark?

- In-memory algorithm
- Matrix versions cached in memory



### Progress

- Revising all linear algebra concepts
- Getting familiar with Scala and Spark
- Trying examples in Python

