MATLAB EXPO 2018

Big Data with MATLAB and Spark

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Real-World Example: Sports Analytics







- Too much data to handle and capture it
- Difficult to predict
- Real-Time dependence



Big data workflow: from desktop to production





So, what's the big (data) challenges?

Standard tools won't work



• Time-consuming



Need to learn new tools & rewrite algorithms





Solution!

Standard tools won't work

Time-consuming

 Need to learn new tools & rewrite algorithms



Prototype algorithms quickly

Run directly from MATLAB with tall arrays

Use the same MATLAB code



Datastore & tall arrays



Use datastore to define file-list
 > ds = datastore('*.csv')

- 2. Create tall table from datastore
 >> tt = tall(ds)
- 3. Act like ordinary table in parallel
 >> model = fitlm(tt.Temp=...)
- 4. Request on local machine
 >> result = gather(tt.result)



Tall arrays: very small changes

1 file

Access Data

measured = readtable('PumpData.csv'); measured = table2timetable(measured);

Preprocess Data

Select data of interest

measured = measured(timerange(seconds(1),seconds(2)),:)

Work with missing data

measured = fillmissing(measured, 'linear');

Calculate statistics

m = mean(measured.Speed);

s = std(measured.Speed);



1000+ files

Access Data

measured = datastore('PumpData*.csv'); measured = tall(measured); measured = table2timetable(measured);

Preprocess Data

Select data of interest

measured = measured(timerange(seconds(1),seconds(2)),:)

Work with missing data

measured = fillmissing(measured, 'linear');

Calculate statistics

m = mean(measured.Speed);

s = std(measured.Speed);

[m,s] = gather(m,s);

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Workflow Pattern

Access out of memory data

Work with subsets of your data

Develop functions for event detection and calculation

Apply functions to all of your data

Aggregate, summarize, & visualize

datastore & tall

findgroups, splitapply

Normal MATLAB code

cellfun

table, histogram, heatmap, boxplot, binScatterPlot



MATLAB Distributed Computing Server (MDCS)

Local Parallel Computing



Deployed Parallel Computing



What is Hadoop/Spark?





Scaling with Spark: Very small changes too!

Desktop Code



Spark + Hadoop Code

Define the Execution Environment

mapreducer(gcp);

Access Data

```
measured = datastore('PumpData*.csv');
measured = tall(measured);
```

Define the Execution Environment

setenv('HADOOP_HOME', '/path/to/hadoop/install')
setenv('SPARK_HOME', '/path/to/spark/install');
cluster = parallel.cluster.Hadoop;

mapreducer(cluster);

Access Data

measured = datastore('PumpData*.csv'); measured = tall(measured);



Big Data with MATLAB & Spark





The MathWorks Fleet Data







Example Setup at MathWorks









Access & Explore Data: MATLAB & Spark MathWorks Vehicle Fleet

- Challenge Develop and deploy Data Analytics to run on Spark against vehicle fleet data stored on Hadoop
- Solution Use MATLAB tall arrays to develop analytics on the desktop and then scale out to the Spark cluster
- ResultsDeveloped insight and understanding of over 1300 vehicle tripsFuel efficiency performance under real-world driving conditions



Analysis Domains

Statistics

- Summary Statistics
- Regression, ANOVA, Machine Learning





Signal Processing

- Sound quality analysis
- LIDAR analysis





Image Processing

Active Safety





Location/Mapping

- Analyzing GPS Data
- Custom Visualizations







Key Takeaways

- Use the **same MATLAB code**
- Use new MATLAB data types datastore & tall arrays for out of memory data sets
- Scale your work up with Parallel Computing Toolbox on the desktop or the MATLAB Distributed Computing Server (MDCS) on Spark





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