SparkAI-Bench: A Benchmark to Generate AI Workloads on Spark

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Outline

- Background and Motivation
- SparkAIbench
  - Overview
  - Process of Workload Generation
  - Available AI Algorithms
  - Expression of Workload Generation Requirement
- Use Case
- Conclusion
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AI workloads in the Cloud

Recent years, distributed machine (deep) learning workloads, referred to as AI workloads, are rapidly becoming prevalent and potential applications in cloud computing.
Existing problems

- There is a lack of workload in the field of artificial intelligence.
  - The major efforts on generating workloads today do not focus on AI domain. And there is no study which is able to automatically generate user customized AI workloads.
  - Workloads generation is one of the most important aspect in benchmarking, generating in a manual manner is quite complicated.

- Example
  - DRL-based scheduler mostly trains agent through the cluster traces generated by running workloads whose characteristics are configured manually due to the lack of frameworks that enable generating diverse and customized user workloads automatically.
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Overview

This paper we present a benchmark to generate AI workloads, which supports a variety of AI algorithms, changeable input data size, as well as parametric method for submission.
Overview

The contributions

- A user customized and automatic AI workloads generator
- A use case to illustrate how SparkAIbench works in a real job scheduling optimization scenario.
SparkAIbench

Process of Workload Generation

1. reading a requirement of AI workloads generation from a JSON file, SparkAIbench is able to know how many workloads should be generated.
SparkAIBench

- 2. select specific machine learning algorithms within Spark MLlib or BigDL according to value of “algorithms”
- 3. according to selected algorithms and the value of “data_size”, SparkAIBench chooses corresponding data generation methods to obtain the training data sets and send them into HDFS.
SparkAIBench

4. package the above algorithms into an assembly jar and put it into YARN-based Spark platform as an application
# SparkAIBench

## Available AI Algorithm

<table>
<thead>
<tr>
<th>Framework</th>
<th>Algorithm</th>
<th>Domain</th>
<th>Data generator</th>
</tr>
</thead>
<tbody>
<tr>
<td>Spark MLlib</td>
<td>Linear Regression</td>
<td>Regression Analysis</td>
<td>SELF</td>
</tr>
<tr>
<td></td>
<td>LDA</td>
<td>Text Mining</td>
<td>SELF</td>
</tr>
<tr>
<td></td>
<td>Bayes</td>
<td>Classification</td>
<td>LibSVM</td>
</tr>
<tr>
<td></td>
<td>SVM</td>
<td>Classification</td>
<td>LibSVM</td>
</tr>
<tr>
<td></td>
<td>FP-Growth</td>
<td>Frequent Itemset Mining</td>
<td>BDGS</td>
</tr>
<tr>
<td></td>
<td>k-means</td>
<td>Clustering</td>
<td>BDGS</td>
</tr>
<tr>
<td></td>
<td>ALS</td>
<td>Recommendation</td>
<td>BDGS</td>
</tr>
<tr>
<td>BigDL</td>
<td>LeNet</td>
<td>Image Classification</td>
<td>MINST</td>
</tr>
<tr>
<td></td>
<td>Inception</td>
<td>Image Classification</td>
<td>ImageNet</td>
</tr>
<tr>
<td></td>
<td>VGG</td>
<td>Image Classification</td>
<td>CIFAR-10</td>
</tr>
<tr>
<td></td>
<td>ResNet</td>
<td>Image Classification</td>
<td>CIFAR-10</td>
</tr>
<tr>
<td></td>
<td>RNN</td>
<td>Natural Language Processing</td>
<td>Tiny Shakespeare Texts corpus</td>
</tr>
<tr>
<td></td>
<td>Auto-encoder</td>
<td>Image Classification</td>
<td>MINST</td>
</tr>
</tbody>
</table>
Workload Generation Requirement

In order to flexibly and controllably represent a user requirement of AI workloads generation, we transform it into a JSON object with several configurable parameters shown in Table (i.e. keys of such JSON object), and insert the object into a JSON file.

<table>
<thead>
<tr>
<th>Name</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>interval</td>
<td>Submission interval among different algorithms.</td>
</tr>
<tr>
<td>algorithms</td>
<td>An integer array with a size of 13 represents which algorithms are selected.</td>
</tr>
<tr>
<td>data_size</td>
<td>An integer array with a size of 13 represents the data scale of each selected algorithm.</td>
</tr>
<tr>
<td>queue</td>
<td>The job queue where the algorithms are submitted to.</td>
</tr>
<tr>
<td>priority</td>
<td>The job priority of each algorithm.</td>
</tr>
</tbody>
</table>
Use Case

- a DRL-based job scheduling optimizer
  - the aim of SparkAIBench in this scenario is to generate various AI workloads for training the job scheduling optimizer (agent).
Use Case

Reward Estimator

The estimator is regarded as a reward function used in DRL mechanism. If carrying out a scheduling decision makes a lower average job latency, it means the scheduling decision improves cluster’s performance, and vice versa.
Use Case

Job Scheduling Optimizer (Agent)

In DRL-based optimizer (agent), two neural networks are introduced, which both take expected accumulated reward as output and with the same model structure.
Use Case

- Proposing Requirements of AI Workloads Generation

**SERVICES**

Generate AI workloads you want

- Input the queue's name: queueA
- Priority: 5
- Interval(s): 15

**Algorithms selection and Data size setting**
- 1000
- LDA
- Bayes
- 2
- FP-Growth
- 1500
- ALS
- LeNet
- Inception
- VGG
- ResNet
- RNN
- Auto-encoder
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Conclusion

- SparkAI Bench
  - a user customized benchmark, SparkAI Bench, with the ability of generating various AI workloads through a configurable user requirement file.

- Project Homepage
  - User manual: https://harryandlina.github.io/
Thanks

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