## **Endterm Report: Distributed LSTM Training & System Monitoring**

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#### Introduction

This report outlines the step-by-step process of integrating Prometheus and Grafana to monitor an Apache Spark distributed data processing system. The setup collects and visualizes key metrics, including CPU load, memory usage, task execution time, and processed data volume. The report includes detailed explanations of each step, with placeholders for screenshots.

- → Preprocessing the ETTh1 dataset (Electricity Transformer Data)
- → Distributing LSTM training across multiple Spark workers
- → Monitoring system performance using Prometheus & Grafana

### **Setting Up Prometheus**

#### **Install Prometheus**

Download Prometheus from prometheus.io.

Open C:\Users\zhanik\Downloads\prometheus\prometheus.yml and add the following:

start prometheus

PS C:\Users\zhanik\Downloads\prometheus-3.2.0-rc.1.windows-amd64> ./prometheus --config.file=prometheus.yml

## **Configuring Spark for Prometheus Metrics**

modify the metrics.properties file:

```
Prometheus Sink for Spark Metrics
 .sink.prometheus.class=org.apache.spark.metrics.sink.PrometheusServlet
*.sink.prometheus.path=/metrics/prometheus
*.sink.prometheus.period=10
#JMX Monitoring for Spark Metrics
*.sink.jmx.class=org.apache.spark.metrics.sink.JmxSink
executor.source.jvm.class=org.apache.spark.metrics.source.JvmSource
executor.sink.jmx.class=org.apache.spark.metrics.sink.JmxSink
executor.sink.prometheusServlet.class=org.apache.spark.metrics.sink.PrometheusServlet
executor.sink.prometheusServlet.path=/metrics/executor
executor.sink.prometheusServlet.port=8082
# JVM Metrics
master.source.jvm.class=org.apache.spark.metrics.source.JvmSource
worker.source.jvm.class=org.apache.spark.metrics.source.JvmSource
driver.source.jvm.class=org.apache.spark.metrics.source.JvmSource
executor.source.jvm.class=org.apache.spark.metrics.source.JvmSource
# Task & Execution Metrics
executor.source.task.class=org.apache.spark.metrics.source.TaskMetricsSource
driver.source.task.class=org.apache.spark.metrics.source.TaskMetricsSource
application.source.task.class=org.apache.spark.metrics.source.TaskMetricsSource
executor.source.blockManager.class=org.apache.spark.metrics.source.BlockManagerSource
driver.source.blockManager.class=org.apache.spark.metrics.source.BlockManagerSource
application.source.blockManager.class=org.apache.spark.metrics.source.BlockManagerSource
```

#### prometheus



### **Enable JMX Exporter for Spark**

Download jmx\_prometheus\_javaagent-1.1.0.jar from GitHub.

Create a jmx\_config.yaml file:

```
rules:
    - pattern: "org.apache.spark<name=.*><>.*"
    - pattern: ".*"
```

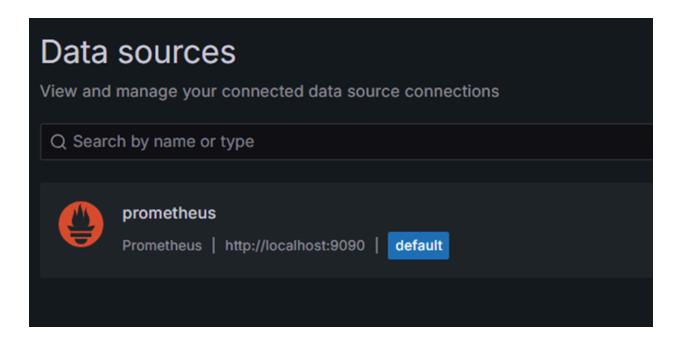
#### **Start Spark with Prometheus and JMX Exporter**

run the command

```
PS C:\Users\zhanik> spark-submit --master spark://192.168.224.224:7077 \
--conf "spark.executor.instances=1" \
--conf "spark.executor.cores=1" \
--conf "spark.executor.cores=1" \
--conf "spark.executor.cores=1" \
--conf "spark.executor.memory=500m" \
--conf "spark.executor.memory=500m" \
--conf "spark.executog.enabled=true" \
--conf "spark.eventLog.enabled=true" \
--conf "spark.eventLog.dir=file:///C:/Users/zhanik/Downloads/spark-3.5.4-bin-hadoop3/logs \
--conf "spark.history.fs.logDirectory=file:///C:/Users/zhanik/Downloads/spark-3.5.4-bin-hadoop3/spark-3.5.4-bin-hadoop3/logs" \
--conf "spark.ui.retainedJobs=50" \
--conf "spark.ui.retainedJobs=50" \
--conf "spark.ui.retainedJasks=50" \
--conf "spark.ui.retainedJasks=50" \
--conf "spark.ui.retainedJasks=50" \
--conf "spark.sql.shirfle.partitions=2" \
--conf "spark.sql.shirfle.partitions=2" \
--conf "spark.default.parallelism=4" \
--conf "spark.executor.processTreeMetrics.enabled=true" \
--conf "spark.executor.processTreeMetrics.enabled=true" \
--conf "spark.driver.extraJavaOptions=-Dcom.sun.management.jmxremote -Dcom.sun.management.jmxremote.port=7078 -Dcom.sun.management.jmxremote.authenticate=false -Dcom.sun.management.jmxremote.ssl=false -javaagent:C:/Users/zhanik/Downloads/spark-3.5.4-bin-hadoop3/spark-3.5.4-bin-hadoop3/jmx_prometheus_javaagent-1.1.0.jar=8078:C:/Users/zhanik/Downloads/spark-3.5.4-bin-hadoop3/spark-3.5.4-bin-hadoop3/jmx_config.yaml" \
--conf "spark.executor.extraJavaOptions=-Dcom.sun.management.jmxremote -Dcom.sun.management.jmxremote.port=7079 -Dcom.sun.management.jmxremote.authenticate=false -Dcom.sun.management.jmxremote.ssl=false -javaagent:C:/Users/zhanik/Downloads/spark-3.5.4-bin-hadoop3/spark-3.5.4-bin-hadoop3/jmx_config.yaml" \
--conf "spark.executor.extraJavaOptions=-Dcom.sun.management.jmxremote.ssl=false -javaagent:C:/Users/zhanik/Downloads/spark-3.5.4-bin-hadoop3/spark-3.5.4-bin-hadoop3/jmx_config.yaml" \
--conf "spark.executor.extraJavaOptions=-Dcom.sun.management.jmxremote.ssl=false -javaagent:C:/Users/zhanik/Downloads/spark-3.
```

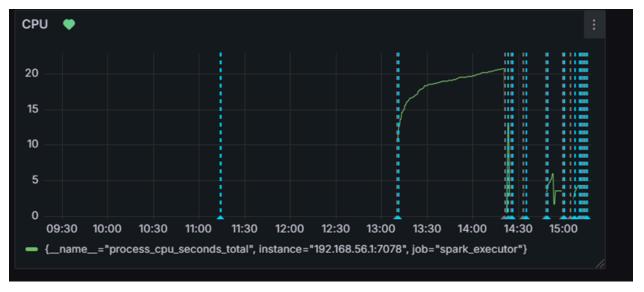
## **Setting Up Grafana**

connect with prometheus

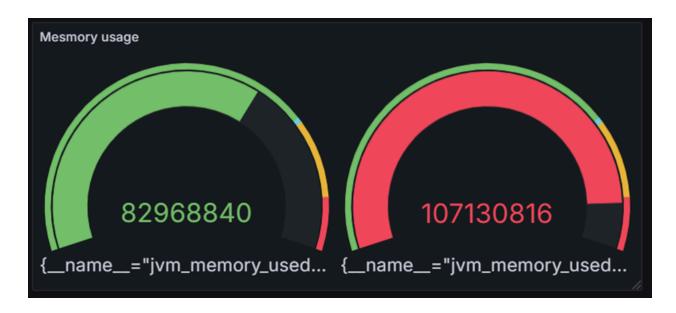


create panels such as

## **CPU Load Panel**



**Memory Usage Panel** 



# Code

This code loads time-series data, preprocesses it, and trains a distributed LSTM neural network using TensorFlow and Apache Spark to predict future values, then saves the trained model.

```
# -*- coding: utf-8 -*-
"""code2.ipynb
Automatically generated by Colab.
Original file is located at
     https://colab.research.google.com/drive/1GgWScTLtJ3BA_NTcnIuZMnjoRbwkmQGh
import os
import pandas as pd
import numpy as np
import tensorflow as tf
from pyspark.sql import SparkSession
from sklearn.preprocessing import MinMaxScaler
os.environ["JAVA_HOME"] = "C:\Program Files\Java\jdk-17" os.environ["SPARK_HOME"] = "C:\\Users\\zhanik\\Downloads\\spark-3.5.4-bin-hadoop3\\spark-3.5.4-bin-hadoop3" spark = SparkSession.builder \
      appName("Distributed LSTM Training v2") \
.ampName("Distributed LSTM Training v2") \
.config("spark://192.168.224.224:7077") \
.config("spark.executor.cores", "1") \
.config("spark.executor.memory", "500m") \
.config("spark.metrics.conf", "metrics.properties") \
      .getOrCreate()
df = spark.read.csv(r"C:\Users\zhanik\ETTh1.csv", header=True, inferSchema=True)
df_pandas = df.select("date", "OT").toPandas()
scaler = MinMaxScaler()
df_pandas["OT"] = scaler.fit_transform(df_pandas["OT"].values.reshape(-1, 1))
def create_sequences(data, seq_length):
      sequences, labels = [], []
for i in range(len(data) - seq_length):
    sequences.append(data[i:i+seq_length])
    labels.append(data[i+seq_length])
      return np.array(sequences), np.array(labels)
seq_length = 24 # Predict next hour using 24 past hours
X, y = create_sequences(df_pandas["OT"].values, seq_length)
df_train = spark.createDataFrame([(x.tolist(), float(y)) for x, y in zip(X, y)], ["sequence", "label"])
num partitions = 4
df_train = df_train.repartition(num_partitions)
strategy = tf.distribute.MultiWorkerMirroredStrategy()
with strategy.scope():
    model = tf.keras.Sequential([
           tf.keras.layers.LSTM(64, return_sequences=True, input_shape=(seq_length, 1)), tf.keras.layers.LSTM(64),
           tf.keras.layers.Dense(32, activation='relu'),
tf.keras.layers.Dropout(0.5),
tf.keras.layers.Dense(1) # Regression output
      1)
      model.compile(optimizer='adam', loss='mse', metrics=['mae'])
def to_numpy_array(df, column):
    return np.array(df.select(column).rdd.flatMap(lambda x: x).collect())
train_data = np.array(to_numpy_array(df_train, "sequence"))
train_labels = np.array(to_numpy_array(df_train, "label"))
train_data = train_data.reshape(train_data.shape[0], train_data.shape[1], 1)
model.fit(train_data, train_labels, epochs=10, batch_size=64)
model.save("/content/distributed_etth1_model.keras")
print(" Distributed LSTM training complete. Model saved.")
```

# spark master



#### - Workers (5)

Worker Id	Address	State	Cores	Memory	Resources
worker-20250214154809-192.168.224.224-52007	192.168.224.224:52007	DEAD	1 (0 Used)	512.0 MiB (0.0 B Used)	
worker-20250214154835-127.0.0.1-49341	127.0.0.1:49341	DEAD	1 (0 Used)	512.0 MiB (0.0 B Used)	
worker-20250214154851-192.168.224.192-49765	192.168.224.192:49765	ALIVE	1 (1 Used)	512.0 MiB (500.0 MiB Used)	
worker-20250214154918-192.168.224.224-52142	192.168.224.224:52142	ALIVE	1 (1 Used)	512.0 MiB (500.0 MiB Used)	
worker-20250214155026-192.168.56.1-53140	192.168.56.1:53140	ALIVE	1 (1 Used)	512.0 MiB (500.0 MiB Used)	

#### - Running Applications (1)

Application ID	Name	Cores	Memory per Executor	Resources Per Executor	Submitted Time	User	State	Duration
app-20250214154930-0001 (kill	Distributed IMD8 Training	3	500.0 MiB		2025/02/14 15:49:30	zhanik	RUNNING	56 s

#### - Completed Applications (1)

Application ID	Name	Cores	Memory per Executor	Resources Per Executor	Submitted Time	User	State	Duration
app-20250214154812-0000	Distributed IMDB Training	1	500.0 MiB		2025/02/14 15:48:12	zhanik	FINISHED	29 s

# spark application dashboard

## Spark Jobs (?)

User: zhanik Total Uptime: 1.6 min Scheduling Mode: FAIR Completed Jobs: 4

- Completed Jobs (4)

Page: 1	43005(4)	1 Pages. Jump to 1 . Show 100 items in a page. Go			
Job Id •	Description	Submitted	Duration	Stages: Succeeded/Total	Tasks (for all stages): Succeeded/Total
3	collect at C\Users\zhanik\code.py;50 collect at C\Users\zhanik\code.py;50	2025/02/14 15:49:55	1s	1/1 (1 skipped)	1/1 (2 skipped)
2	csv at NativeMethodAccessorImpl.javax0 csv at NativeMethodAccessorImpl.javax0	2025/02/14 15:49:51	4 s	1/1 (1 skipped)	1/1 (2 skipped)
1	showString at NativeMethodAccessorImpl.javac0 showString at NativeMethodAccessorImpl.javac0	2025/02/14 15:49:45	4 s	1/1 (1 skipped)	4/4 (2 skipped)
0	runJob at PythonRDD.scala:181 runJob at PythonRDD.scala:181	2025/02/14 15:49:38	4 s	2/2	3/3
Page: 1					1 Pages. Jump to 1 . Show 100 items in a page. Go