資工碩一 113598007 楊明哲

環境設置

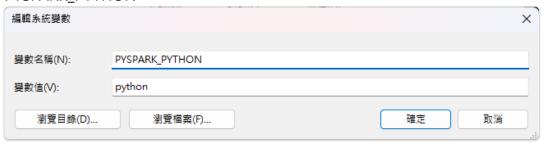
1. Windows

• 環境版本

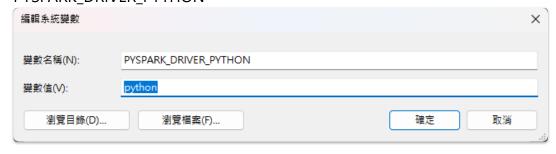
Java: 17.0.12Python: 3.10.10PySpark: 3.5.3

• Pandas: 2.2.3

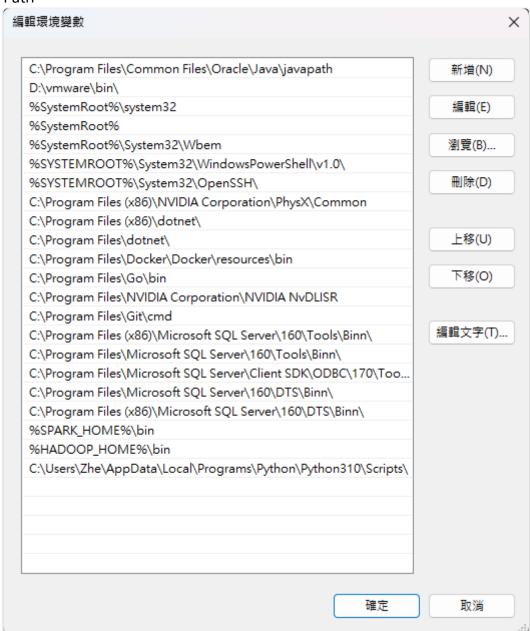
- 環境配置
 - 下載指定版本的 Python 和 Java
 - 透過 pip install pyspark 安裝 pyspark · 並且將它初始化
 - 將下載的 winutils.exe 放進 pyspark 中 bin 的資料夾下
 - 設置環境變數
 - PYSPARK_PYTHON



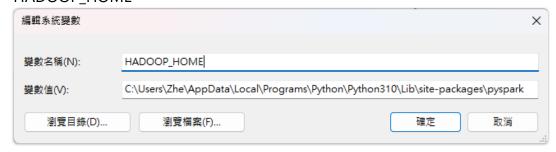
PYSPARK_DRIVER_PYTHON



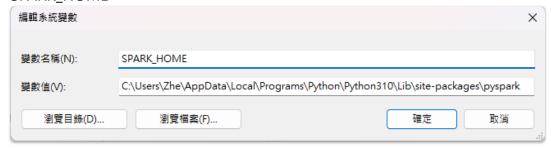
Path



HADOOP_HOME



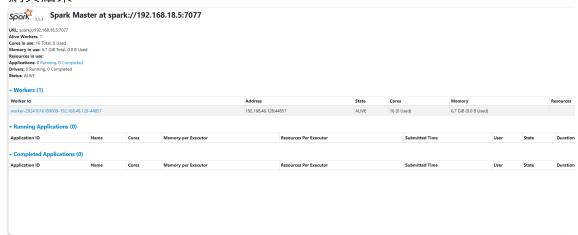
SPARK_HOME



- 執行指令 spark-class org.apache.spark.deploy.master.Master,並打開 URL
 - 指令結果

```
FS D:\Code\Big|S Data\Mining> spark-class org.apache.spark.deploy.master.Master
Using Spark's default logaj profile: org/apache/spark/logaj2-defaults.properties
24/19/16 23:33:15 IMFO Master: Started daemon with process name: 12148@CESKTOP-F8USCHM
24/19/16 23:33:15 IMFO Master: Started daemon with process name: 12148@CESKTOP-F8USCHM
24/19/16 23:33:16 IMFO SecurityManager: Changing view acls sro: Zhe
24/19/16 23:33:16 IMFO SecurityManager: Changing modify acls to: Zhe
24/19/16 23:33:16 IMFO SecurityManager: Changing modify acls to: Zhe
24/19/16/16 23:33:16 IMFO SecurityManager: Changing modify acls to: Zhe
24/19/16/16 23:33:16 IMFO SecurityManager: Changing modify acls groups to:
24/19/16/16 23:33:16 IMFO SecurityManager: SecurityManager: aclt method idealled; ui acls disabled; users with view permissions: Zhe; groups with modify permissions: EMPTY
24/19/16/16 23:33:16 IMFO Master: Starting Spark master at spark://192.168.18.5:78077
24/19/16 23:33:16 IMFO Master: Starting Spark wersion 3.5.3
24/19/16 23:33:16 IMFO Master: Running Spark wersion 3.5.3
24/19/16 23:33:16 IMFO Master: Running Spark wersion 3.5.3
24/19/16 23:33:16 IMFO Master: Running Spark wersion 3.5.3
24/19/16 23:33:16 IMFO Master: Starting Spark wersion 3.5.3
24/19/16 23:33:16 IMFO Master: Starting Spark wersion 3.5.3
24/19/16 23:33:16 IMFO Master: Imply Started service 'MasterUI
24/19/16 23:33:16 IMFO Master: Running Spark version 3.6.9, and started at http://DESKTOP-F8USCHM:8080
24/19/16 23:33:16 IMFO Master: I have been elected leader! New starte: ALIVE
24/19/16 23:33:16 IMFO Master: Registering worker 192.168.46.128:44857 with 16 cores, 6.7 GIB RAM
```

• 網頁結果



2. Ubuntu

環境版本

Memory : 8GB

Processors: 16

Java: 17.0.12

Python: 3.10.12

• Spark: 3.5.3

• 環境配置

• 下載指定的 Java 版本

• 下載指定的 Spark,並且解壓縮到 Documents 下

• 將各環境變數設置

```
cluster0@cluster0-virtual-machine: ~
 export JAVA_HOME=/usr/lib/jvm/java-17-openjdk-amd64
export SPARK_HOME=/home/cluster0/Documents/spark-3.5.3-bin-hadoop3
 export HADOOP_HOME=/home/cluster0/Documents/spark-3.5.3-bin-hadoop3
export PATH=$PATH:$SPARK HOME/bin:$HADOOP HOME/sbin
 ~/.bashrc: executed by bash(1) for non-login shells.
 see /usr/share/doc/bash/examples/startup-files (in the package bash-doc)
 for examples
 If not running interactively, don't do anything
 ase $- in
    *i*) ;;
f don't put duplicate lines or lines starting with space in the history.
# See bash(1) for more options
HISTCONTROL=ignoreboth
# append to the history file, don't overwrite it
shopt -s histappend
                                                                16,4
                                                                              Тор
```

 執行 spark-class org.apache.spark.deply.worker.Worker spark://192.168.18.5:7077

網頁書面



並且能夠在 Master 上的網頁看到 Worker 的資訊



源代碼

```
import pandas as pd
from pyspark.sql import SparkSession
  from pyspark.sql.functions import col, min, max, mean, stddev
  spark = SparkSession.builder.appName("Electric Power Consumption").getOrCreate()
 # Read Data
file_path = "D:/Code/Big_Data_Mining/Electric_Power_Consumption/household_power_consumption.txt"
  df = pd.read_csv(file_path, sep=';', low_memory=False)
 # Drop missing values
df.dropna(inplace=True)
# Data type conversion

df['Global_active_power'] = df['Global_active_power'].astype(float)

df['Global_reactive_power'] = df['Global_reactive_power'].astype(float)

df['Voltage'] = df['Voltage'].astype(float)

df['Global_intensity'] = df['Global_intensity'].astype(float)
 # Convert pandas DataFrame to Spark DataFrame
spark_df = spark.createDataFrame(df)
min_values = spark_df.select(
min(col('Global_active_power')).alias('Global_active_power'),
min(col('Global_reactive_power')).alias('Global_reactive_power'),
min(col('Voltage')).alias('Voltage'),
min(col('Global_intensity')).alias('Global_intensity')
  ).collect()[0]
         _values = spair_ur.select(
max(col('Global_active_power')).alias('Global_active_power'),
max(col('Global_reactive_power')).alias('Global_reactive_power'),
max(col('Global_intensity')).alias('Global_intensity')
max(col('Global_intensity')).alias('Global_intensity')
  count values = spark df.count()
  stats1 = pd.DataFrame({
    'Min': [smi_values['Global_active_power'], min_values['Global_reactive_power'], min_values['Voltage'], min_values['Global_intensity']],
    'Min': [max_values['Global_active_power'], max_values['Global_reactive_power'], max_values['Voltage'], max_values['Global_intensity']],
    'Count': [count_values, count_values, count_values, count_values]
}, indexe['Global_active_power', 'Global_reactive_power', 'Voltage', 'Global_intensity'])
  print("Q1:")
print(stats1)
  # Q2: Find mean, standard deviation
mean_values = spark_df.select(
        an_values = spark_df.select(
mean(col('Global_active_power')).alias('Global_active_power'),
mean(col('Global_reactive_power')).alias('Global_reactive_power
mean(col('Voltage')).alias('Voltage'),
mean(col('Global_intensity')).alias('Global_intensity')
  ).collect()[0]
 studev_values = spark_df.select(
    stddev(col('Global_active_power')).alias('Global_active_power'),
    stddev(col('Global_reactive_power')).alias('Global_reactive_power'),
    stddev(col('Voltage')).alias('Voltage'),
    stddev(col('Global_intensity')).alias('Global_intensity')
).collect()[0]
 stats2 = pd.DataFrame({
    'Mean': [mean_values['Global_active_power'], mean_values['Global_reactive_power'], mean_values['Global_intensity']],
    'StdDev': [stddev_values['Global_active_power'], stddev_values['Global_reactive_power'], stddev_values['Global_intensity']]
}, index=['Global_active_power', 'Global_reactive_power', 'Voltage', 'Global_intensity'])
  # Q3: Min-Max Normalization
normalized_df = spark_df.select(
           ((col('Global_reactive_power') - min_values['Global_reactive_power']) / (max_values['Global_reactive_power'] - min_values['Global_reactive_power'])).alias('Normalized_Global_((col('Voltage') - min_values['Voltage'] - min_values['Voltage']).alias('Normalized_Voltage'),

((col('Global_intensity') - min_values['Global_intensity']) / (max_values['Global_intensity'] - min_values['Global_intensity']).alias('Normalized_Global_intensity'))
  stats_output_path = "D:/Code/Big_Data_Mining/result/Q2.csv"
normalized_output_path = "D:/Code/Big_Data_Mining/result/Q3.csv"
  stats1.to_csv(stats1_output_path)
  stats2.to_csv(stats2_output_path)
normalized_df.toPandas().to_csv(normalized_output_path, index=False)
```

輸出結果

執行 spark-submit --master spark://192.168.18.5:7077 --conf spark.driver.host=192.168.18.5 homework0.py

第一題

Find min, max, count

```
01:
                            Min
                                      Max
                                              Count
Global_active_power
                          0.076
                                   11.122
                                            2049280
Global_reactive_power
                          0.000
                                    1.390
                                           2049280
Voltage
                        223.200
                                  254.150
                                            2049280
Global intensity
                                   48.400
                          0.200
                                            2049280
```

第二題

Find mean, standard deviation

第三題

Normalization

```
\textbf{0.3747963063552418,} \\ \textbf{0.30071942446043165,} \\ \textbf{0.376090468497577,} \\ \textbf{0.37759336099585067} \\ \textbf{0.3747963063552418,} \\ \textbf{0.30071942446043165,} \\ \textbf{0.376090468497577,} \\ \textbf{0.37759336099585067,} \\ \textbf{0.37609046849757,} \\ \textbf{0.3760904684975,} \\ \textbf{0.3760904684975,} \\ \textbf{0.37609046849,} \\ \textbf{0.37609046849,} \\ \textbf{0.37609046849,} \\ \textbf{0.376090468,} \\ \textbf{0.37609046,} \\ \textbf{0.3760904,} \\ \textbf{0.
\textbf{0.4783632084012313,0.31366906474820144,0.33699515347334413,} \\ \textbf{0.47302904564315357} \\ \textbf{0.4730290456431537} \\ \textbf{0.4730290456431} \\ \textbf{0.4730290456431} \\ \textbf{0.4730290456431} \\ \textbf{0.4730290456431} \\ \textbf{0.47302904564} \\ \textbf{0.47302904564} \\ \textbf{0.47302904564} \\ \textbf{0.47302904564} \\ \textbf{0.47302904564} \\ \textbf{0.473029045} \\ \textbf{0.473029045
0.4796306355241717,0.35827338129496406,0.32600969305331173,0.47302904564315357
0.48089806264711216,0.3611510791366907,0.3405492730210021,0.47302904564315357
\textbf{0.3250045265254391,} \textbf{0.3798561151079137,} \textbf{0.4032310177705981,} \textbf{0.32365145228215775}
\textbf{0.311787072243346,} \\ \textbf{0.37553956834532376,} \\ \textbf{0.3819063004846531,} \\ \textbf{0.3070539419087137} \\ \textbf{0.3070539419087} \\ \textbf{0.3070539419087} \\ \textbf{0.307053941} \\ \textbf{0.3070539419087} \\ \textbf{0.30705394} \\ \textbf{0.30705
0.3282636248415716,0.3741007194244605,0.3841680129240713,0.32365145228215775
0.32808256382400874,0.3741007194244605,0.38836833602584825,0.32365145228215775
\textbf{0.32518558754300203}, \textbf{0.36690647482014394}, \textbf{0.3486268174474964}, \textbf{0.32365145228215775}
0.39579938439254037,0.35827338129496406,0.31211631663974215,0.40248962655601667
0.48307079485786714,0.3381294964028777,0.30953150242326355,0.47717842323651455
0.46605105920695283,0.34388489208633094,0.3163166397415191,0.46058091286307057
\textbf{0.470034401593337,0.2863309352517986,0.3137318255250405,} 0.4647302904564316
0.29947492304906753,0.2028776978417266,0.4504038772213244,0.29045643153526973
0.28915444504798116,0.10935251798561152,0.43715670436187376,0.28215767634854777
0.30363932645301467,0.11223021582733814,0.4478190630048467,0.29460580912863077
\textbf{0.28879232301285535,} \\ \textbf{0.0,0.4500807754442649,} \\ \textbf{0.28215767634854777} \\ \textbf{2.28215767634854777} \\ \textbf{2.282157676348547} \\ \textbf{2.28215767637} \\ \textbf{2.28215767637} \\ \textbf{2.2821576767} \\ \textbf{2.2821576767} \\ \textbf{2.28215767} \\ \textbf{2.282157} \\ \textbf{2
0.3306174180698896,0.0,0.4084006462035544,0.3360995850622407
0.5267065000905305,0.0,0.3066235864297255,0.5228215767634855
\textbf{0.690747782002535,} \\ \textbf{0.0,0.25137318255250396,} \\ \textbf{0.6846473029045643}
\textbf{0.6291870360311426,0.0,0.2911147011308567,} 0.6307053941908715
0.4615245337678799,0.0,0.35508885298869153,0.4522821576763486
0.3981531776208583,0.0,0.37996768982229445,0.39834024896265563
0.2871627738547891,0.0,0.43489499192245556,0.27800829875518673
\textbf{0.28607640774941157,0.0,0.4084006462035544,} 0.27800829875518673
0.2853521636791599,0.0,0.4006462035541195,0.27800829875518673
0.28806807894260367,0.0,0.39709208400646245,0.28215767634854777
  0.28082563824008694.0.0.0.390306946688207.0.2738589211618257
```

補充

• 網頁顯示結果

- Completed Applications (1)								
Application ID	Name	Cores	Memory per Executor	Resources Per Executor	Submitted Time	User	State	Duration
app-20241016235426-0000	Electric Power Consumption	16	1024.0 MiB		2024/10/16 23:54:26	Zhe	FINISHED	1.3 min