

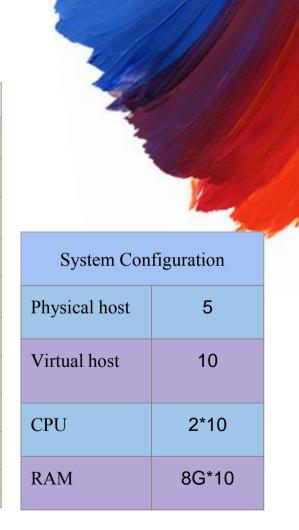
大數據分析平台Hadoop架設

第一組: 戴澤宇、林穎奇、潘敏政、楊為今、蔡汶吟

▶ 此報告由第一組全體同仁協力完成

組員分工表

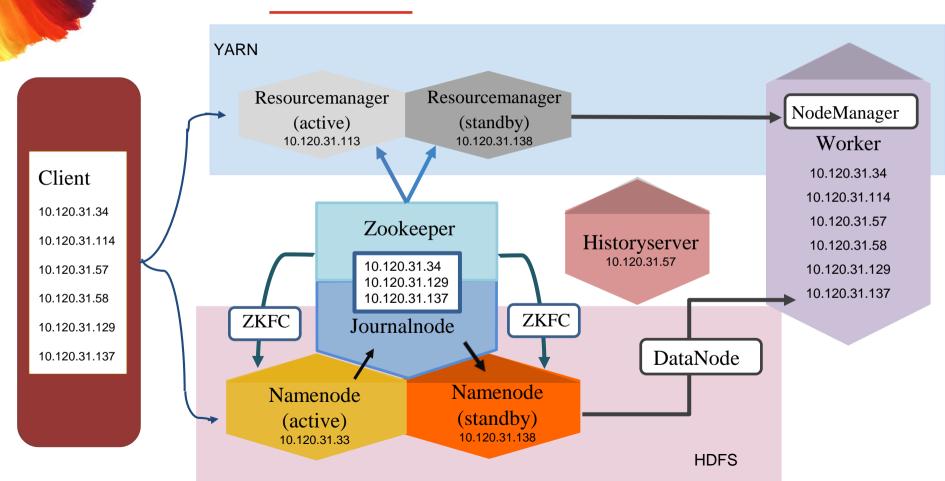
User / Local Host ip	Virtual host	HA clusters
戴澤宇 10.120.31.32	Namenode 10.120.31.33	
	10.120.31.34	Journalnode/Zookeeper
林穎奇 10.120.31.112	Resourcemanager	
	10.120.31.114	
楊為今 10.120.31.56	JobHistory 10.120.31.57	
	10.120.31.58	
蔡汶吟 10.120.31.128	10.120.31.129	Journalnode/Zookeeper
	10.120.31.130	Standby Resource Manager
潘敏政 10.120.31.136	10.120.31.137	Journalnode/Zookeeper
	10.120.31.138	Standby Namenode



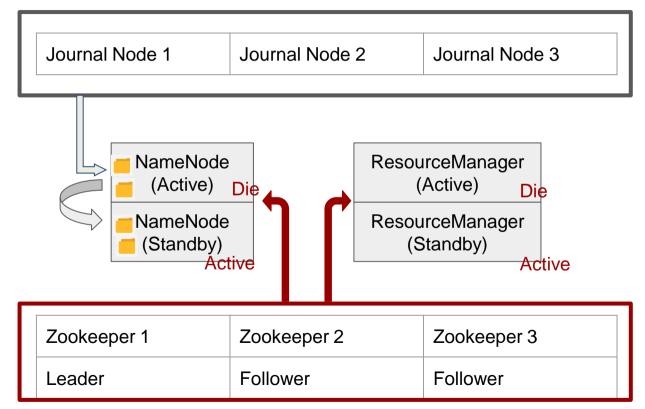
HADOOP 叢集架構節點說明

NameNode (Master Node)	負責記錄與維護HDFS的Metadata Metadata有:檔案名稱・檔案與其區塊(Blocks)的對應關係・各區塊所在的 資料節點(DataNode)・其它的檔案屬性(如:建立時間・複本數量…等)		
ResourceManager	管理叢集系統中所有應用程序的資源分配		
JobHistoryServer	紀錄執行Map-Reduce各作業的花費時間和啟動、結束時間等詳細資料		
DataNode (Slave Node)	檔案區塊(Blocks)實際儲存的地方,通常有多部DataNode DataNode會定期傳送現有的Blocks狀態給NameNode,若NameNode發現某個Block之複本數量少於現有的開份設定時,NameNode會自動增加該Block的複本。 以當某個DataNode掛掉時,NameNode會自動將此DataNode上的所有Blocks重新配置到其它DataNode上		
NodeManager	負責監控各container資源使用情況(cpu、記憶體)・回報給RM		
JournalNode	為了同步兩個NameNode·standby NameNode從這獲取NameNode備份的資料		
Zookeeper	為分佈式應用提供一致性服務的軟件,提供的功能包括:配置維護、域名服務、分佈式同步、組服務等。		

HADOOP.HA



High Availability (Zookeeper & JournalNode)

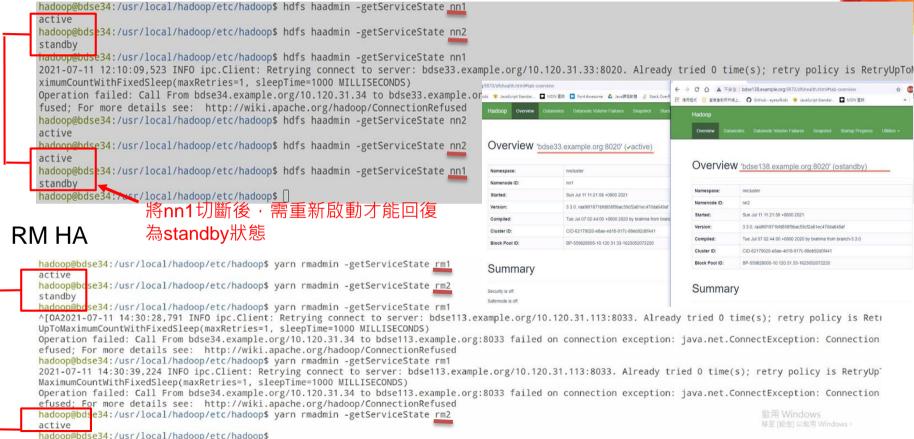


ZKFailoverController 能及時檢測到 NameNode 的健康狀況,在主 NameNode、

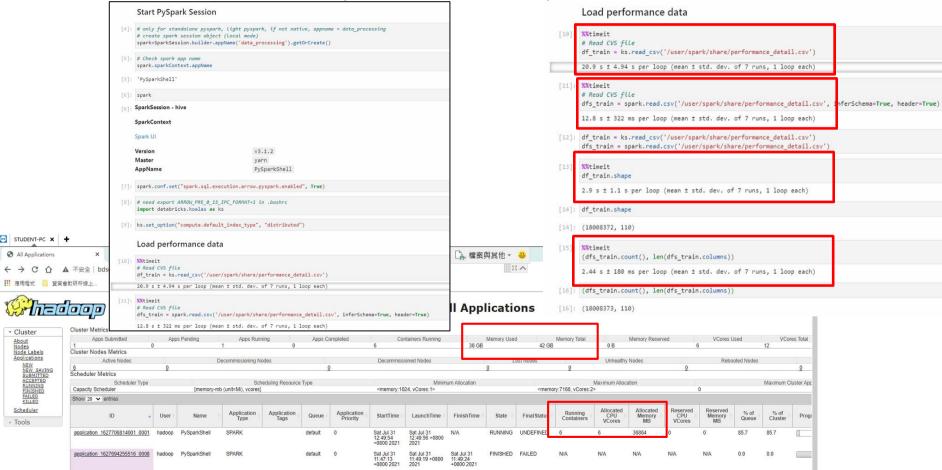
ResourceManager 故障時藉助 Zookeeper 實現自動的主備選舉和切換

NN HA





Run Performance_detail.csv (18008372筆資料)



Troubleshooting

啟動 historyserver, historyserver 自己輸入jps做確認時發現螢幕末顯示出JobHistoryServer

- cd /usr/local/hadoop
- ◆ Is -I 有 logs
- cd /usr/local/hadoop/logs
- ◆ less -N 看 hadoop-hadoop-historyserver-bdsexx.logs 檔案

```
155 rver(JobHistoryServer.java:224)
156 at org.apache.hadoop.mapreduce.v2.hs.JobHistoryServer.main(JobHistorySer
156 ver.java:232)
157 Caused by: java.net.BindException: Port in use: master3.example.org:19888
158 .java:1292)
159 at org.apache.hadoop.http.HttpServer2.bindForSinglePort(HttpServer2.jav
159 :1314)
160 at org.apache.hadoop.http.HttpServer2.openListeners(HttpServer2.java:137
160 3)
161 at org.apache.hadoop.http.HttpServer2.start(HttpServer2.java:1223)
162 at org.apache.hadoop.yarn.webapp.WebApps$Builder.start(WebApps.java:472)
```



Troubleshooting

- cd \$HADOOP_HOME
- cd etc/hadoop
- nano mapred-site.xml

Change master3.example.org:19888 to bdse57.example.org:19888

<figure>

OKAY





測試執行程式

```
[1]: import time
      import numpy as np
      import pandas as pd
      import pyspark
      import sys
     import pyspark.sql.functions as fn
[3]: from pyspark.ml.regression import RandomForestRegressor
      from pyspark,ml.tuning import ParamGridBuilder, CrossValidator
      from pyspark.ml.evaluation import RegressionEvaluator
      from pyspark.ml.feature import VectorAssembler
[8]: import databricks.koalas as ks
     df = ks.read csv('/user/spark/share/AF Stock TW 2603.TW.csv')
     df = df.drop('key 0', axis='columns')
11]: tar = ks.DataFrame(df['close'].shift(periods=20).rename('label'))
[12]: ks.set_option('compute.ops_on_diff_frames', True)
     df = ks.concat([df, tar], axis=1)
131: # 保留前 100 row 空值/\\\\\ 空於 60% 的 column
     cond = df.iloc[:101,:].isnull().sum()/100 <= 0.6
     df = df[cond[cond == True].index.to numpy()]
141: # 看 58 row 以後的空信
     df.iloc[58:,:].isnull().any().sum()
14]: 0
[15]: # 取 58 row 以後當新的 df # index reset
     df = df.iloc[58:,:].reset index(drop=True)
16]: df.shape
16]: (2508, 184)
     sdf = df.to_spark()
```

Build Random Forest Model

```
[25]: rf = RandomForestRegressor()
      rf model = rf.fit(train df)
[26]: # predict on the test set
      model predictions = rf model.transform(test df)
[27]: # print prediction
      model predictions.show(10)
                   features
                                         label
                                                       prediction
       [19.2769391502943...|19.276939392089844| 19.18613926767244
      [19.6624788274294... | 18.660076141357425 | 19.10773054039324
```

[19.8937998275340...|17.619121551513672|19.019794481807274 [20.4335541213312... | 18.852846145629883 | 19.72018766766709 [21,2046327145946...]19,778139114379883]20,249421125173686 [21.2817385819022... | 19.084169387817383 | 19.78803131285374 [21.5901708137132...] 19.35404586791992 20.057975469364557 |(183,[0,1,2,3,4,5...| 19.97090721130371|20.249421125173686 [19.4311537671171...|20.317893981933597| 20.30185747590754 only showing top 10 rows

Evaluate Model

```
[28]: evaluatorRMSE = RegressionEvaluator().etLabelCol('label').setPredictionCol("prediction").setMetricName("rmse")
```

```
[29]: RMSE = evaluatorRMSE.evaluate(model predictions)
      RMSE
```

```
[29]: 1.282201095799959
```

30]:	evaluatorR2	RegressionEvaluator()	setLabelCol('label').setPredictionCol	("prediction").setMetricName("r	2
------	-------------	-----------------------	---------------------------------------	---------------------------------	---

```
481: # FEATURE IMPORTANCES
     best rf model.featureImportances
```

```
48]: SparseVector(183, {0: 0.0, 1: 0.0001, 2: 0.0, 3: 0.0, 4: 0.0003, 5: 0.0043, 6: 0.0001, 7: 0.0007, 8: 0.0, 9: 0.0, 10: 0.0, 11:
              0, 19: 0.0, 20: 0.0, 21: 0.015, 22: 0.0, 23: 0.0, 24: 0.0011, 25: 0.0, 26: 0.0, 27: 0.0049, 29: 0.0, 30: 0.0003, 33: 0.0003, ]
              0, 43: 0.0, 44: 0.0001, 45: 0.0001, 46: 0.0001, 47: 0.0, 48: 0.0007, 49: 0.0002, 50: 0.0001, 51: 0.0002, 52: 0.0002, 53: 0.0,
               0.0, 62: 0.0, 63: 0.0, 64: 0.0, 65: 0.0013, 66: 0.0001, 67: 0.0002, 68: 0.0, 69: 0.0, 70: 0.0, 71: 0.0007, 72: 0.0001, 73: 0.0
              82: 0.0007, 83: 0.0, 84: 0.0, 85: 0.0001, 86: 0.0003, 87: 0.1565, 88: 0.0, 89: 0.045, 90: 0.0, 93: 0.0, 95: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100: 0.0, 100:
              115: 0.0, 116: 0.0, 117: 0.0, 118: 0.0, 119: 0.0, 121: 0.0, 122: 0.0, 124: 0.0, 128: 0.0, 129: 0.0, 130: 0.0, 131: 0.0, 137: (
              0.0002, 153: 0.0, 154: 0.0, 155: 0.0, 156: 0.0014, 157: 0.0003, 158: 0.0, 159: 0.0001, 160: 0.0, 161: 0.0, 162: 0.0, 163: 0.0,
               1: 0.0168, 172: 0.0, 173: 0.0024, 174: 0.0735, 175: 0.0268, 176: 0.0, 177: 0.0001, 178: 0.0001, 179: 0.003, 180: 0.0, 181: 0.0
```

