

Terapixel Project

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Project Requirements

Written Report Outline

What is the need for the project?

Justify your choice of response (i.e. the nature of, and your plan for, your project). To give strength to your argument you should reference to practice elsewhere (e.g. in academic literature, or industry practices)

Discuss and implementation of CRISP-DM

How successful has it been? Provide evidence, using appropriate evaluation methodologies, and comment on the strengths/weaknesses of your evidence in answering this question

What are the future implications for work in this area? If applicable, which areas of extension work are now possible due to the foundational work you have performed in this project?

A brief reflection on your personal and professional learning in undertaking this project. Here, you can comment on how you found the process, what you learned about the technologies and methodologies you used, which aspects you found most difficult/straightforward, and any conclusions which will inform the way you undertake similar projects in the future

Structured Abstract

REMEMBER TO INCLUDE STRUCTURED ABSTRACT

EDA Findings.

You should also produce additional documentation detailing the findings from your exploratory analysis. You are encouraged to make use of a literate programming framework, e.g. R Markdown, to align analytic code with narrative text. You should submit the source file(s) for the notebooks as well as output saved in PDF Format. No limit on length for this document!

CRISP-DM Methodology:

Business Understanding

Determine Business Objectives

Background

Maybe include more background information on Newcastle University and data

With world-leading research in data, Newcastle University is always undertaking new challenges to advance the field. In order to keep up with ever-expanding amount of data produced by cities, one such challenge is to be able to effectively convey the information from this data to stakeholders. To address this issue, the University has created a terapixel image of the city of Newcastle upon Tyne, including environmental data such as temperature and humidity, from sensors across the city. This terapixel image contains over one trillion pixels and not only allows the entire city to be viewed, but is rendered at 12 different levels allowing users to zoom in whilst retaining full picture detail. However, rendering over one trillion pixels requires immense computing power, and so the University has made use of public cloud service *Microsoft Azure* to perform the rendering process. This facilitates a far quicker rendering process, which is more suitable for a primary requirement of the project - to ensure the terapixel image supports daily updates.

Business Objectives

Whilst Newcastle University has created this terapixel visualisation, it is important that the rendering process with cloud supercomputing is rigorously evaluated in order to identify any inefficiencies.

Business Success Criteria

Assess Situation

Determine Data Mining Goals

Produce Project Plan

Data Understanding

Describe Data

In order to evaluate the rendering process, Newcastle University has provided three files detailing various aspects of a single run of the rendering process. The file “application.checkpoints.csv” has dimensions and variable names:

```
dim(application.checkpoints)
```

```
## [1] 660400      6
```

```
colnames(application.checkpoints)
```

```
## [1] "timestamp" "hostname" "eventName" "eventType" "jobId" "taskId"
```

```
str(application.checkpoints)
```

```
## tibble [660,400 x 6] (S3: tbl_df/tbl/data.frame)
## $ timestamp: chr [1:660400] "2018-11-08T07:41:55.921Z" "2018-11-08T07:42:29.842Z" "2018-11-08T07:42:59.842Z" ...
## $ hostname : chr [1:660400] "0d56a730076643d585f77e00d2d8521a00000N" "0d56a730076643d585f77e00d2d8521a00000N" "0d56a730076643d585f77e00d2d8521a00000N" ...
## $ eventName: chr [1:660400] "Tiling" "Saving Config" "Saving Config" "Render" ...
## $ eventType: chr [1:660400] "STOP" "START" "STOP" "START" ...
## $ jobId    : chr [1:660400] "1024-1v112-7e026be3-5fd0-48ee-b7d1-abd61f747705" "1024-1v112-7e026be3-5fd0-48ee-b7d1-abd61f747705" "1024-1v112-7e026be3-5fd0-48ee-b7d1-abd61f747705" ...
## $ taskId   : chr [1:660400] "b47f0263-ba1c-48a7-8d29-4bf021b72043" "20fb9fcf-a927-4a4b-a64c-70258b60c0d0" "20fb9fcf-a927-4a4b-a64c-70258b60c0d0" ...
```

Where the “eventType” can be either “START” or “STOP” to indicate the end of beginning of an event, and “eventName” can take the following values;

```
unique(application.checkpoints$eventName)
```

```
## [1] "Tiling"          "Saving Config" "Render"        "TotalRender"  
## [5] "Uploading"
```

In which “TotalRender” is a combination of all other events, which represent different parts of the whole rendering process. In addition to “eventType” and “eventName”, “timestamp” gives the time associated with this event starting or stopping, “hostname” is the hostname of the virtual machine auto-assigned by the Azure batch system and “jobId” and “taskId” are the unique IDs of the Azure batch job and task, respectively.

The file “gpu.csv” has the following dimensions and variable names, with all variable values being of class character:

```
dim(gpu)
```

```
## [1] 1543681      8
```

```
colnames(gpu)
```

```
## [1] "timestamp"      "hostname"        "gpuSerial"       "gpuUUID"  
## [5] "powerDrawWatt"  "gpuTempC"        "gpuUtilPerc"     "gpuMemUtilPerc"
```

```
str(gpu)
```

```
## tibble [1,543,681 x 8] (S3: tbl_df/tbl/data.frame)  
## $ timestamp      : chr [1:1543681] "2018-11-08T08:27:10.314Z" "2018-11-08T08:27:10.192Z" "2018-11-08T08:27:10.192Z" ...  
## $ hostname       : chr [1:1543681] "8b6a0eebc87b4cb2b0539e81075191b900001C" "d8241877cd994572b46c861" ...  
## $ gpuSerial      : num [1:1543681] 3.23e+11 3.24e+11 3.23e+11 3.25e+11 3.23e+11 ...  
## $ gpuUUID       : chr [1:1543681] "GPU-1d1602dc-f615-a7c7-ab53-fb4a7a479534" "GPU-04a2dea7-f4f1-12d" ...  
## $ powerDrawWatt : num [1:1543681] 131.6 117 121.6 50.2 141.8 ...  
## $ gpuTempC      : int [1:1543681] 48 40 45 38 41 43 41 35 43 36 ...  
## $ gpuUtilPerc   : int [1:1543681] 92 92 91 90 90 88 91 0 93 90 ...  
## $ gpuMemUtilPerc: int [1:1543681] 53 48 44 43 47 40 47 0 56 40 ...
```

The variable names “timestamp” and “hostname” have the same meaning as in “application.checkpoints.csv”. “gpuSerial” and “gpuUUID” are the serial number and unique ID of the physical GPU card, which represent a one-to-one match the “hostname” variable. As there are 1024 gpu cores, there are 1024 different values for “hostname”, “gpuSerial” and “gpuUUID”.

```
length(unique(gpu$hostname))
```

```
## [1] 1024
```

```
length(unique(gpu$gpuSerial))
```

```
## [1] 1024
```

```
length(unique(gpu$gpuUUID))
```

```
## [1] 1024
```

```
dim(unique(gpu[c("hostname", "gpuSerial", "gpuUUID")]))
```

```
## [1] 1024    3
```

The additional variables in this data set are quantitative variables describing different aspects of the cores. These include “powerDrawWatt” (the power draw of the GPU in watts), “gpuTempC” (the temperature of the GPU in Celsius), “gpuUtilPerc” (the percent utilisation of the GPU core) and “gpuMemUtilPerc” (the percent utilisation of the GPU memory).

We can also observe that variables “timestamp”, “hostname” and “gpuUUID” are of class character, whilst variables “gpuSerial” and “powerDrawWatt” are of class numeric. Lastly, variables “gpuTempC”, “gpuUtilPerc” and “gpuMemUtilPerc” are integer values.

```
str(gpu)
```

```
## tibble [1,543,681 x 8] (S3: tbl_df/tbl/data.frame)
## $ timestamp      : chr [1:1543681] "2018-11-08T08:27:10.314Z" "2018-11-08T08:27:10.192Z" "2018-11-08T08:27:10.192Z" ...
## $ hostname       : chr [1:1543681] "8b6a0eebc87b4cb2b0539e81075191b900001C" "d8241877cd994572b46c861" ...
## $ gpuSerial       : num [1:1543681] 3.23e+11 3.24e+11 3.23e+11 3.25e+11 3.23e+11 ...
## $ gpuUUID         : chr [1:1543681] "GPU-1d1602dc-f615-a7c7-ab53-fb4a7a479534" "GPU-04a2dea7-f4f1-12d" ...
## $ powerDrawWatt   : num [1:1543681] 131.6 117 121.6 50.2 141.8 ...
## $ gpuTempC        : int [1:1543681] 48 40 45 38 41 43 41 35 43 36 ...
## $ gpuUtilPerc     : int [1:1543681] 92 92 91 90 90 88 91 0 93 90 ...
## $ gpuMemUtilPerc  : int [1:1543681] 53 48 44 43 47 40 47 0 56 40 ...
```

The final file that we will use is “task.x.y.csv”. This file has the dimensions and variables:

```
dim(task.x.y)
```

```
## [1] 65793    5
```

```
colnames(task.x.y)
```

```
## [1] "taskId" "jobId" "x"      "y"      "level"
```

Variables “jobId” and “taskId” are the same as in “application.checkpoints.csv”, whilst “x” and “y” represent the location of each pixel being rendered. The “Level” variable allows users to zoom into the visualisation. In total there are 12 levels, however only levels 12, 8 and 4 are rendered whilst the other levels are derived in the tiling process. Level 12 is when the image is at maximum zoom, and level 1 is when the image is zoomed right out. The “jobId” relates to the “Level” variable, so in total there are three different “jobId” values corresponding to the three levels that are rendered, as shown below:

```
unique(task.x.y[c("jobId", "level")])
```

```
## # A tibble: 3 x 2
##   jobId                                level
##   <chr>                                <int>
## 1 1024-lvl12-7e026be3-5fd0-48ee-b7d1-abd61f747705    12
## 2 1024-lvl4-90b0c947-dcfc-4eea-a1ee-efe843b698df      4
## 3 1024-lvl8-5ad819e1-fbf2-42e0-8f16-a3baca825a63      8
```

We can also observe that the values in “taskId” and “jobId” are of class character, whilst the other variables are of class integer.

```
str(task.x.y)
```

```
## tibble [65,793 x 5] (S3: tbl_df/tbl/data.frame)
## $ taskId: chr [1:65793] "00004e77-304c-4fbd-88a1-1346ef947567" "0002afb5-d05e-4da9-bd53-7b6dc19ea6d"
## $ jobId : chr [1:65793] "1024-lvl12-7e026be3-5fd0-48ee-b7d1-abd61f747705" "1024-lvl12-7e026be3-5fd0"
## $ x      : int [1:65793] 116 142 142 235 171 179 255 218 241 22 ...
## $ y      : int [1:65793] 178 190 86 11 53 226 61 250 166 220 ...
## $ level  : int [1:65793] 12 12 12 12 12 12 12 12 12 12 ...
```

Data Quality

With the data described at a high level, it is necessary to assess the quality of it. First we can check for any missing values using the below code.

```
sum(is.na(application.checkpoints))
```

```
## [1] 0
```

```
sum(is.na(gpu))
```

```
## [1] 0
```

```
sum(is.na(task.x.y))
```

```
## [1] 0
```

As we are satisfied that there is no missing data, we can now check the data for duplicates.

```
dim(application.checkpoints)[1] - dim(unique(application.checkpoints))[1]
```

```
## [1] 2470
```

```
dim(gpu)[1] - dim(unique(gpu))[1]
```

```
## [1] 9
```

```
dim(task.x.y)[1] - dim(unique(task.x.y))[1]
```

```
## [1] 0
```

We can see that there are 2,470 duplicates in “application.checkpoints.csv”, 9 in “gpu.csv” and none in “task.x.y.csv”. With no missing data and only a small percentage of the data in duplicate, the data is of good quality and we can proceed with the data preparation phase of the analysis.

Look for blanks

Look for duplicates

Data Preparation

Go through process of creating full merged dataset to combine all three

Modelling

Create several graphs to view different interactions of data set

Which event types dominate task runtimes? (app checkpoints)

What is the interplay between GPU temperature and performance? (gpu)

What is the interplay between increased power draw and render time? – Can we quantify the variation in computation requirements for particular tiles?

Can we identify particular GPU cards (based on their serial numbers) whose performance differs to other cards? (i.e. perpetually slow cards).

What can we learn about the efficiency of the task scheduling process?

Evaluation

Deployment