

Advertisement detection system using machine learning techniques.

Final Project Report

DT211c

BSc in Computer Science (Infrastructure)

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Abstract

This project revolves around the detection of advertisements using machine learning means, creating a system that can apply these means and the evaluation of that system. The motivation behind this system is the turmoil in advertisement deliverance. The use of advertisements to track users as they navigate the internet and data privacy concerns that arise because of this. Further research into this space has revealed advertisement blockers use the same technology to complete their task. New methods of completing this task should be developed.

This report describes how this system was designed and how the design will be implemented. The backend will utilise two main machine learning algorithms, word2vec and the inception model. The system will be used through a chrome plugin implemented in JavaScript. The middle layer will then connect these algorithms and front-end.

The final system will be a chrome plugin that highlights advertisements on a website and presents a metric for how sure the system is.

Declaration

I hereby declare that the work described in this dissertation is, except where otherwise stated, entirely my own work and has not been submitted as an exercise for a degree at this or any other university.

Signed:

\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_\_

Student Name

Date

Acknowledgements

I would like to acknowledge the support I have received during the undertaking of this project.

My supervisor, Brian Gillespie, for guiding this project.

My family for their ability to translate my English into readable English.

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# 1. Introduction

## Project Background

In January of 2019, Google proposed a change in their extension manifest version (v3) (1). The point of this update is to make extensions safer by adding greater permission controls. This however had the knock-on effect of “killing” ad block support according to Raymond Hill, the maintainer of uBlock origin and uMatrix (2). A conclusion therefore can be made that extensions cannot be used reliably to protect privacy while browser creators have a vested interest in creating targeted ads.

On the 25th of May 2018, the EUs General Data Protection Regulation (GDPR)(3) came into effect. GDPR has made it mandatory for websites to ask for consent to use the user’s data for personalizing content and ads. The problem here is often if you do not agree, you are unable to access the content on the website. In other countries the user is not given the option and consent is implied by the user using the service. This makes ad blocking necessary to protect the privacy of users who want to use such websites.

Observations were made of the turmoil between advertisement suppliers and advertisement blockers, as a result research to find other solutions to the Ad blocking problem commenced. Pi-hole, a DNS sinkhole for advertisements appeared as a promising solution. This program is simple to install but requires an always on server. The problem with this solution is it required prior knowledge of computing whereas browser plugins are a 1-click install meaning they are far simpler for the greater public to use.

## Project Description

To solve the problems mentioned above, a solution was proposed where the system uses a convolutional neural network model to Identify images in a website. The output will then be compared against what the text on the page is about. To compare the text and image classification output, word2vec a word embedding 2-layer neural network will be used decide what the text is about. If there is a miss match, we can assume that the suspect image is an advertisement. As time goes on advertisement providers are getting craftier or have the ability to remove the impact of the ad blocking software, as a result more ads are slipping through the cracks. This system would get around these issues by not having static ad lists that must be maintained.

Do a diagram

## Project Aims and Objectives

The goal of this project is to decide if a website has adverts or not. To achieve this objective the images and text on the page will be classified. The outcome of these two machine learning algorithms will then be compared against each other. This will be done by grouping these words based on what they are. For example, a dog is in the canine category were as classes will be in an item category. These are two separate groups, so it is likely the image is an advertisement. In this project, the concept will be refined and tested as a valid method of detection.

To achieve this a number of objectives must be completed. Step one is to gather relevant technology’s and test their suitability for the systems application.

## Project Scope

This project is about ad detection. Ad removal is out of scope for this project, the system will inform the user on how likely an image is or is not an advertisement. For the purpose of this system, the scope will be narrowed to the type of advertisements and text the system will deal with. This can be expanded depending on how accurate the system is at predicting the likely hood of the image being an advertisement. Starting with a small number of advertisement types will allow testing and validate if this is a relevant method of checking for advertisements.

As an application the system must be lightweight and fast. The interface should follow the concepts of universal design.

## Thesis Roadmap

One sentence summary of the following chapters This section will provide a summary of each of the chapters covered in this report.

*Research*

This chapter delves into existing methods of ad detection. Following this, an examination into current/existing technologies and data that will enable the completion of this project. Finally, this chapter will discuss any other research done for this project.

*Design*

This chapter explains the design and overall layout of the system. An explanation each layer of the system will be explained. This will include UML diagrams.

*Development*

The development chapter goes through the entire development process of the prototype system. The middle-tier and backend systems that where discussed in the Design and Research chapters are implemented here.

*Testing and Evaluation*

In this chapter I discuss the testing methodologies used and define the parameters for success.

*Redevelopment*

This chapter goes onto explain what steps must be taken to incorporate the technologies used in the prototype into the final system.

*Conclusions and Future Work*

This chapter reflects on the completed prototype system and research completed. The conclusions drawn from this work will be used to propel the development of this system.

# 2. Literature Review

## 2.1. Introduction

In this chapter, key technologies to the success of the system will be discussed. A Comparison to alternative existing solutions to this problem will be made.

## 2.2. Alternative Existing Solutions to Your Problem

*uBlock Origin*

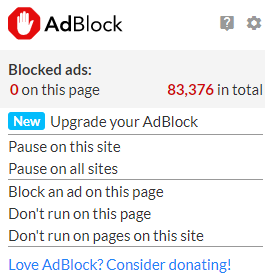
An efficient blocker: easy on memory and CPU footprint, and yet can load and enforce thousands more filters than other popular blockers out there (4). This is a complex interface for an app of this type. This is because the domains of each ad that are blocked are displayed. Control is limited to an on and off button. The app also has html element removing functionality.



*Figure 1- uBlock origin interface*

*AdBlock*

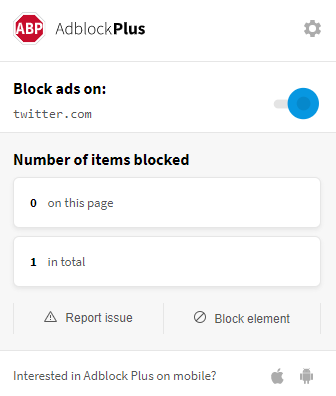
AdBlock (5) does exactly what it says on the tin. It supports the mainstream web browsers as well as android and iOS. The user interface displays the least information but allows the users to pause block and control running in one simple page.



*Figure 2- AdBlock interface*

*AdBlock Plus*

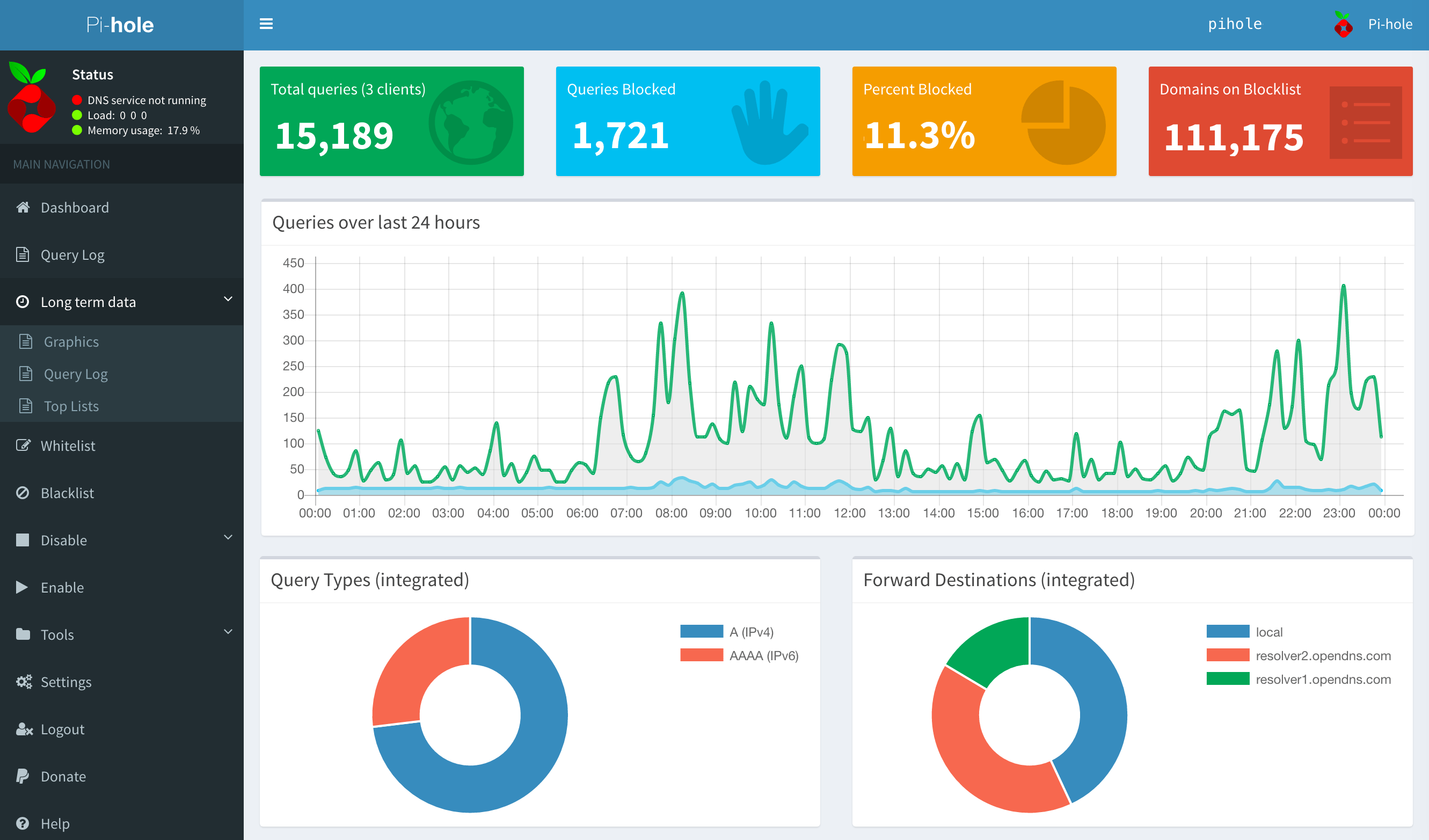
AdBlock plus (6) provides a distraction free browsing experience in the same was as others in this list. AdBlock plus is open-source and gives users access to privacy controls. The interface is very simple, this is a toggle for blocking ads on the current website and a display for advertisements blocked.



*Figure 3- AdBlock Plus*

*Pi-hole*

Pi-hole (7) is a DNS sinkhole that is used to provide network wide adblocking. This is another system that uses static lists to find and block advertisements. The system is harder to setup then its counterparts as it is hosted on an internal server and client DNS setting must be changed to point to the internal server. The system can also double as a DHCP server.



*Figure 4- Pi-hole dashboard*

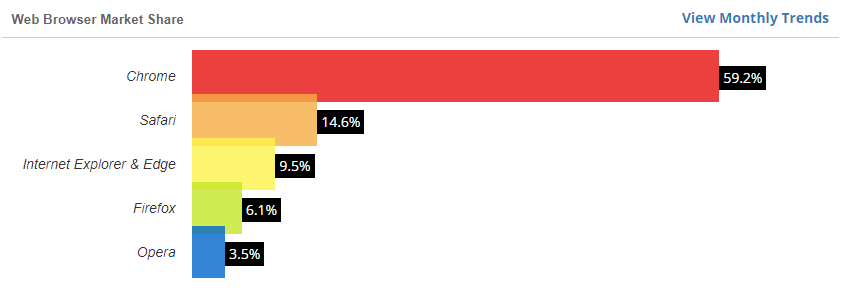
*Conclusion*

These programs all use a variation of static lists to find adverts. The most common are plugins for common browsers i.e. Firefox, chrome, etc. All these solutions have very simple interfaces that allow the user to see the total number of advertisements blocked on the site. They also allow the user to enable and disable on the fly. All the plugin mentioned have 10,000,000+ users on google chrome making this a very sought-after functionality.

## 2.3. Technologies you’ve researched

The following are possible platforms:

*Chrome*

Chrome is a popular web browser (8). Chrome currently has about 60 percent of the browser market according to statcounter and w3counter. These statistics are collected from all websites that use W3Counters or statcounter, these websites collect telemetry using JavaScript allowing insights into the website’s performance. Google are a major supplier of advertisements, so have a vested interest in crippling the current systems (9). 

*Figure 5- W3counter report*

*Firefox*

Firefox (10) does not have anywhere near the popularity of chrome. Firefox has a reputation for user privacy and in 2019 have rolled out features such as a VPN and have removed plugins that collect excessive user data. By default, 3rd party tracking cookies and crypto miners are blocked. Firefox is backed by a non-profit.

*Conclusion*

In conclusion, chrome is the market leader for web browsers. A method that does not uses static lists will be most useful on this platform as this is where the advertisement skirmish is happening. Firefox actively supports user privacy meaning that if the system is successful it should be ported to this platform in the future.

The following are possible programming languages:

*Java script*

This is the go-to language for extension development along with html and CSS. Java script can be used to modify the DOM on websites this will be used to highlight or completely remove the ads.

*Python*

Python is a high-level interpreted language that emphasised code reliability. Python supports many open source frameworks, library’s and development tools. Python also allows for rapid prototyping due to the small size of the code. Applications like conda or virtualenv allow for python programs to be isolated from other projects and help manage their dependencies.

*C++*

C++ is a general-purpose compiled language. C++ is used mainly in systems programming and embed systems because of its features and security. C++ has support for library’s and frameworks.

*Conclusion*

In conclusion, python’s ability to produce rapid prototypes and mature library’s make it a good language for this system. Python is often used in machine learning and web development tasks.

The following are possible library’s and models:

*Word2vec*

This is a method of converting words to a machine-readable format. Normally to deal with categorical features like words one must embed them into a numerical format but in doing so it loses how the words relate to each other. Word2vec preserves the relationship between the words. Word2vec comes in two flavours, the skip-gram method of word2vec weighs nearby context words more heavily than more distant context words and Bag of words which the order of context words does not influence prediction. (11)

*Inception*

Inception is a convolutional network that classifies images. Unlike its competitors it goes wider instead of deeper by having multiple convolutions on the same level. Using a pre trained model as opposed to building a new model makes sense when trying to solve this problem as these models are optimized and trained well past anything one person can manage on their own. (12)

*WordNet*

WordNet® is a large lexical database of English. Nouns, verbs, adjectives and adverbs are grouped into sets of cognitive synonyms (synsets), each expressing a distinct concept. Synsets are interlinked by means of conceptual-semantic and lexical relations. (13) The linking of the items in the wordnet database and our ability to see what Synset the word is part of will allow the system to make meaningful comparisons between the output of the convolutional neural network and text analysis systems

TensorFlow Lite

This allows TensorFlow models on mobile, embedded, and IoT devices (14). Lite does this by using pre-trained models. These models are designed to execute efficiently on mobile and other embedded devices with limited compute and memory resources. TensorFlow Lite has a python implementation this should allow me to use python as the backend of my extension/website. There is also a JavaScript implementation, this could be used to run the image r

The following are possible web frameworks:

*Flask and flask restful*

Flask (15) is a micro web framework for python that provide the means to create a web application. It is free and open source. Flask restful builds on this functionality allowing developers to create REST APIs. A rest API may be necessary for this system depending on what portions of this project can be ran in browser. Flask is a micro framework meaning that functionality like logging in and Restful APIs are added in separate libraries or must be implemented by the developer.

*Django*

Django is a web framework for python that includes all the functionality needed to scale and secure a website. Django like flask is free and open source but is more feature rich out of the box.

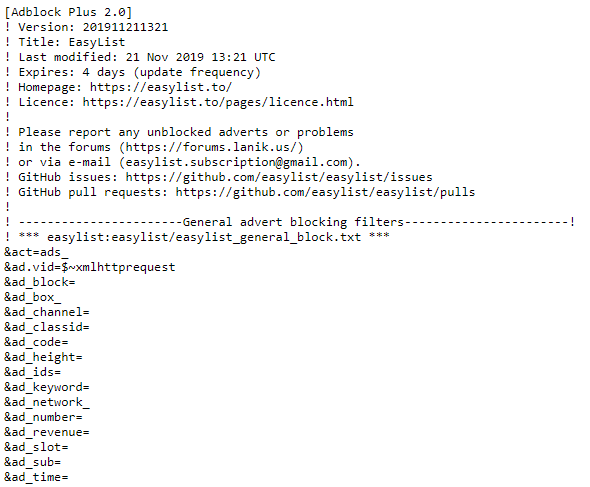
*Conclusion*

Flask is a far simpler and more flexible as it sets out to be a mini web framework and nothing more. Whereas Django aims to be an all-inclusive platform. As of now I only require a method of communicating between the middle layer and backend because of this the lighter simpler platform will be better for this project.

## 2.4. Other Research you’ve done

*Ad lists*

These are lists made and maintained by the AdBlock community. A popular Ad list, EasyList (16) comes pre-packed and/or tested with many adblockers (AdBlock Plus, AdBlock and uBlock Origin). EasyList is a primary filter list for removing most adverts from international webpages, including unwanted frames, images and objects. It is the basis for many other filter lists.



## 2.5. Existing Final Year Projects

Here are some final year projects from previous years that are relevant to this project. These projects are both machine learning projects and as such share some similarities to this project in their development style, technologies and methods used.

*Title:* Euro Coin Classification. Using Image Processing & Machine Learning

*Student:* Yumin Chen

*Description (brief):*

This project aims to investigate the visual features of euro coins and develop models that can represent different natural images. The objective of this project is to investigate the use of machine learning specifically in the context of euro coin classification.

*What is complex in this project*

Classifying the coin is the difficult part of the project. This is done using both image processing and machine learning. The student implements different algorithms and evaluates them.

*What technical architecture was used*

OpenCV, Cordova, machine learning and data mining techniques.

*Explain key strengths and weaknesses of this project, as you see it.*

The student tested a multitude of ml algorithms and evaluated them against each other. I see this as a strength. The student uses Rapid Application Development methodology, prototypes were made quickly but the problem of immediately starting to code is an issue. This weakness in the development style, extra care should be taken in the planning stage for this sort of project.

*Title:* Detecting Bot Twitter Accounts using Machine Learning

*Student:* Emmet Hanratty

*Description (brief):*

In this project the student creates a system to detect and evaluate whether a user account is a bot. The system manages good accuracy, and this will be increased as the data set expands. They system also creates graphs on the data used in prediction.

*What is complex in this project*

Training a sentiment model based on tweet data. This decides if a tweet is positive, natural or negative.

*What technical architecture was used*

Python using the Django framework, MySQL, Twitter API and Git

*Explain key strengths and weaknesses of this project, as you see it*.

The project is not multi-threaded meaning it is slow in the pre-processing stages. Twitter have a limit on the amount of calls you can make to their API, meaning after a certain number of requests the system is unusable for the rest of the day.

The system has good accuracy when predicting the state of an account and this accuracy will grow over time.

2.6. Conclusions

From looking at the current popular solutions for advertisement blocking we can see they all employ a similar approach for this task. With certain companies having a vested interest in curtailing this technology for their own profit and said companies having the means of doing so other solutions should be developed. To develop a solution to this problem possible technologies were researched, and existing Final year projects were looked at to provide guidance.

# 3. Experiment Design

## 3.1 Introduction

In this chapter,

## 3.2. Software Methodology

*Waterfall Methodology*

The waterfall model (17) is a phase by phase method of developing software. The current phase must be completed before commencing the next phase. Once a phase is finished it cannot be revisited till the end of the project.



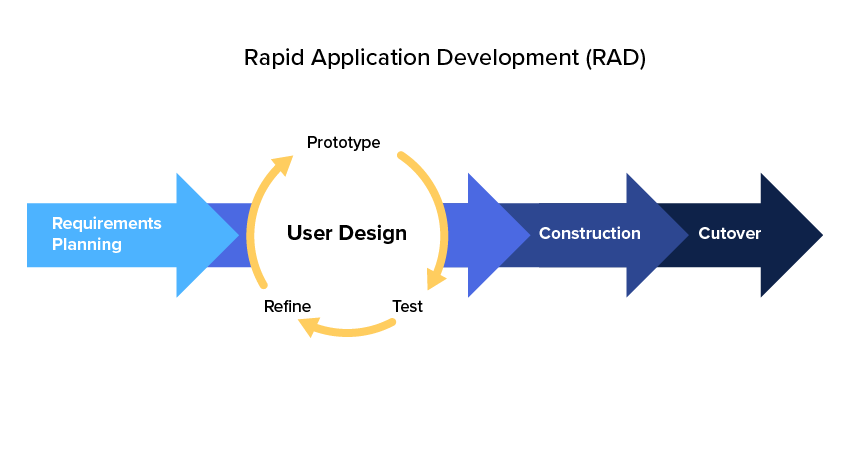
*Figure 7- Waterfall model*

An advantage of this model includes clear and understandable milestones. This makes the stages of the process and advancements very clear.

A disadvantage of the waterfall model is that is difficult to redevelop and redesign when the goalpost changes. Once started you are locked into this mode of development till the maintenance phase. On completion of the project, the application may or may not be suitable for its purpose.

*Rapid application development*

Rapid application development (18) is a form of agile software development. This methodology will allow rapid testing and creation of new approaches.



*Figure 8- Rapid Application Development*

The main advantage of the RAD methodology is that time between prototypes and their next iteration is short. This allows for rapid development over a shorter period.

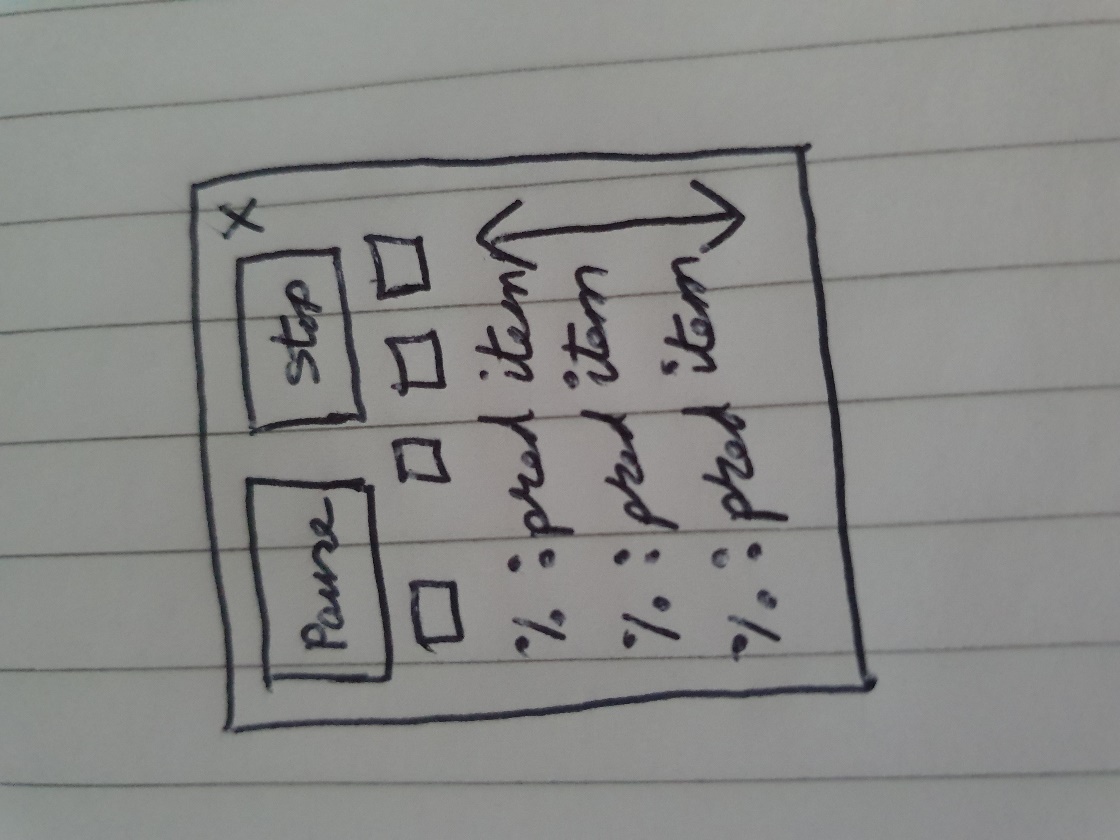
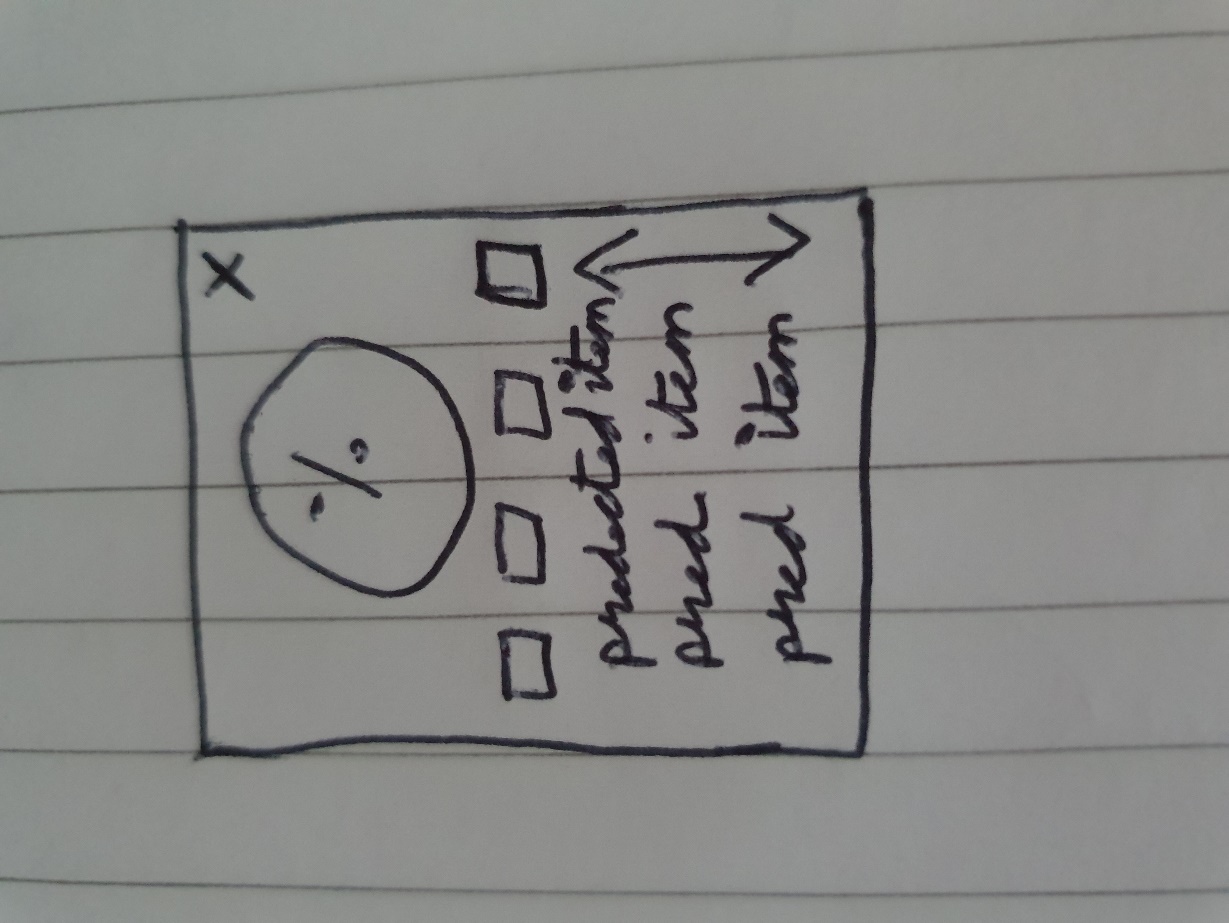
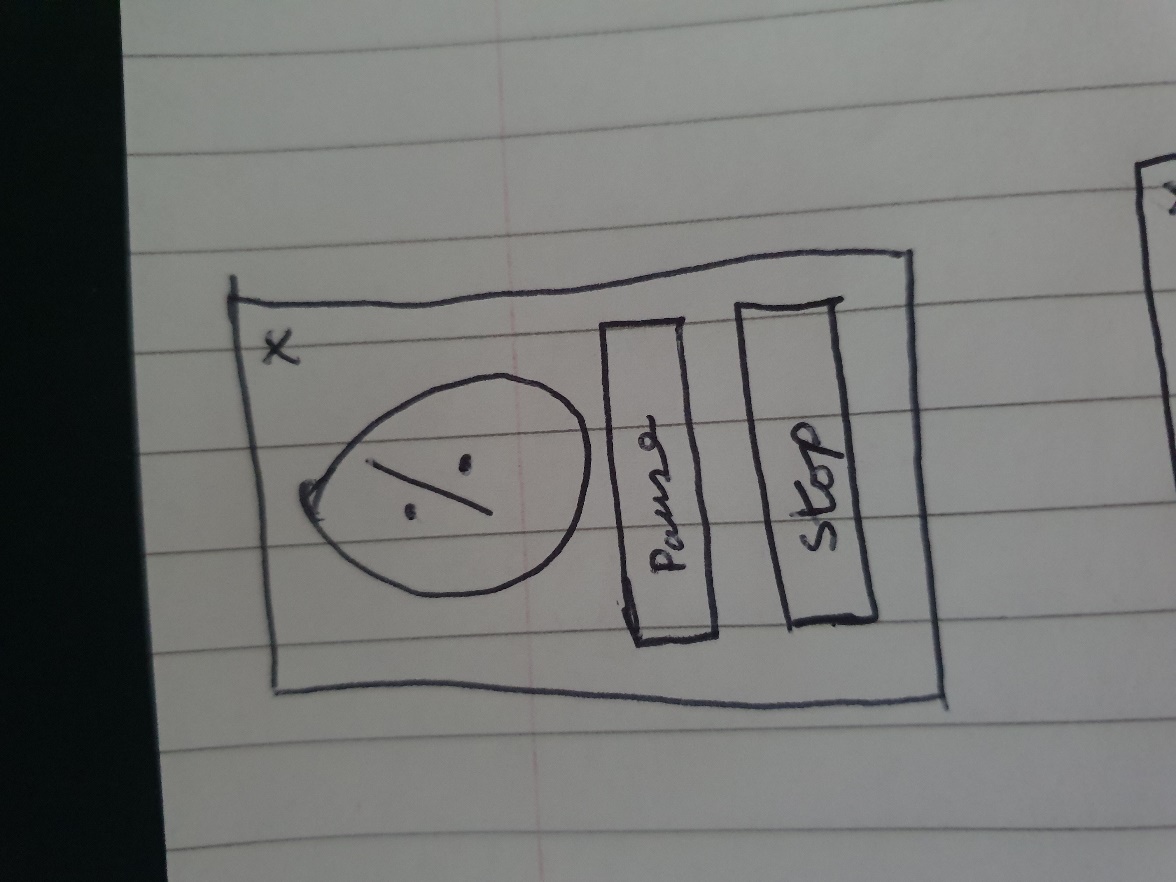
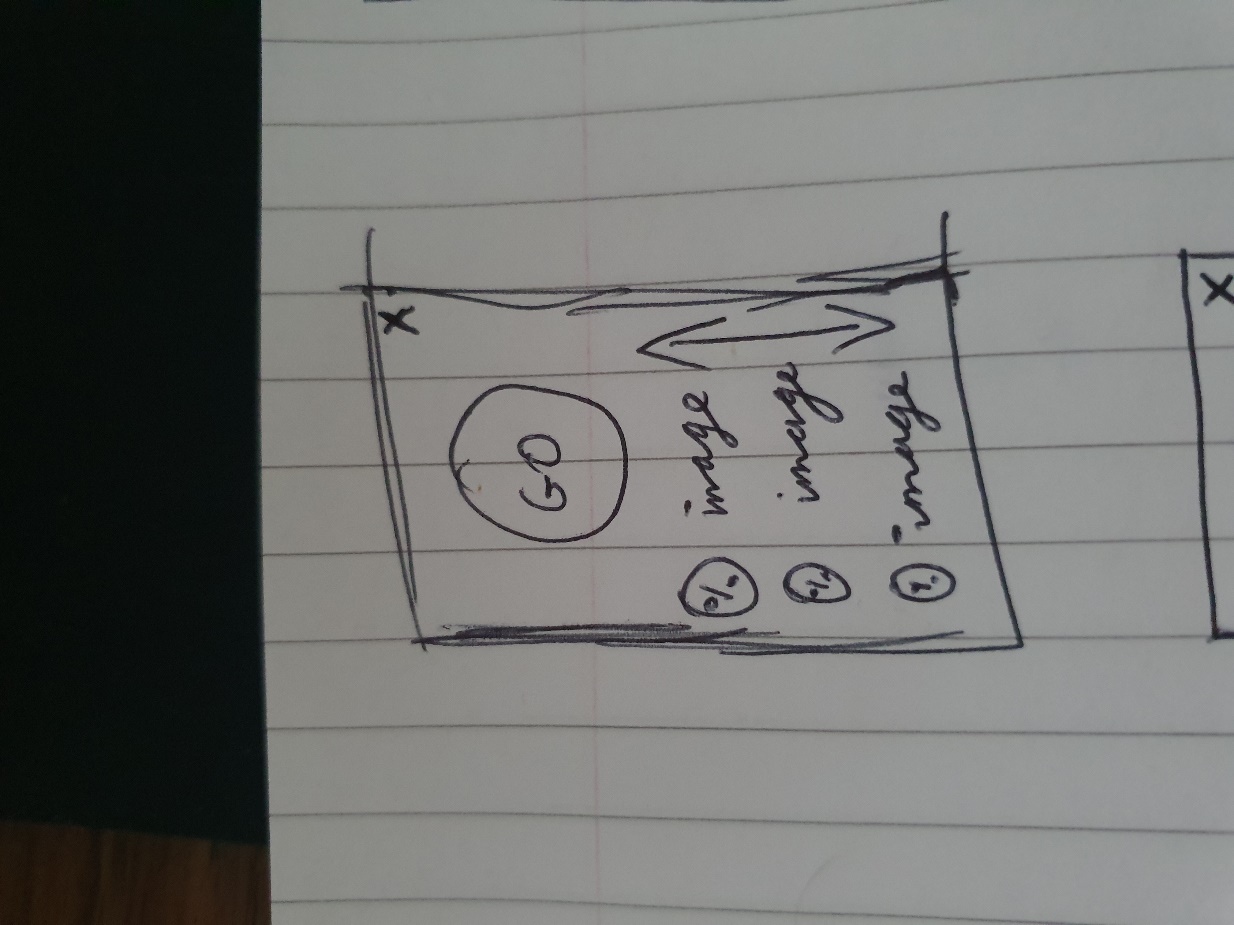
The disadvantages of RAD are mainly based around applying the methodology to larger teams. As this is a solo project this is not a problem. Another disadvantage of this methodology is it is not suitable for larger projects.

## 3.3. Overview of System

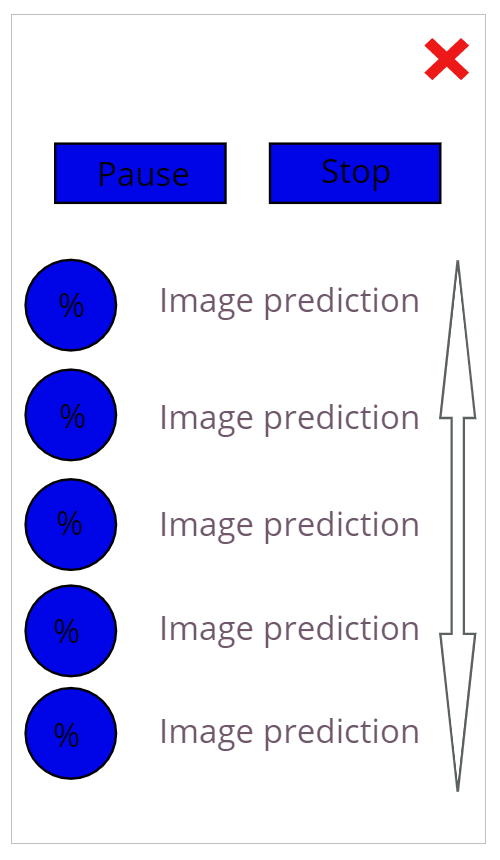
Using the Rapid application development method of software design the first step is define what the system aims to accomplish, this has been completed in part 1.3 of this report. The second stage of RAD is to prototype designs and develop these prototypes. The general approach to the prototype phase was to divide the system into manageable pieces. These pieces were then built, refined and tested. After Unit testing which will serve as the feedback, these pieces will be grouped together to create the final system.

## 3.4. Front-End

The system will have a similar front end to its alternative solutions. This will be written in html, CSS and JavaScript as these are the default languages used to write extensions for chrome. Across these solutions, the number of ads blocked, a way to stop and restart the application are universal. Below are possible designs for the interface of the application.



The next step was to create medium fidelity prototypes. The front end will we simple with one to two different pages. The prototype displays the functionality of the interface and its overall layout. A simple easy to understand interface will server this system the best as it will need to be accessible to all types of users.



The final design will.

The front end is also responsible for the parsing and gathering of the data. This is then posted to the Flask API.

## 3.5. Middle-Tier

This is the flask API responsible for interacting with the backend systems and the front end.

The API will utilise both get and post requests. The get requests will be responsible for explaining to the user what and how the API should be used. Posting to the API on different routes will allow the chrome extension to interact with the backend systems. This separation will allow lower spec computers to take advantage of these algorithms.

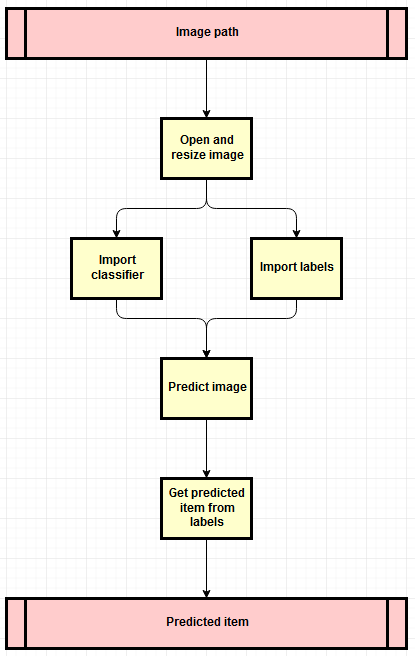
## 3.6. Back-End

The backed will consist of 3 systems, image classification, word embeddings and .

*Image classifier*

The image classification system will be based on a retrained inception mode inception. This model is already trained on ImageNet making its predictions already quite accurate, retraining this model for even more common items that we see in advertisements will give an even better result. Training for logos is another option letting us specialize the classifier even more.

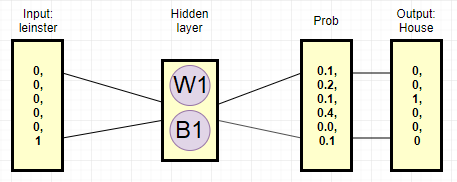
Below is a possible structure for the image prediction. In the final version this will run in browser using TF lite.



*Figure 12 - Image Classification Design*

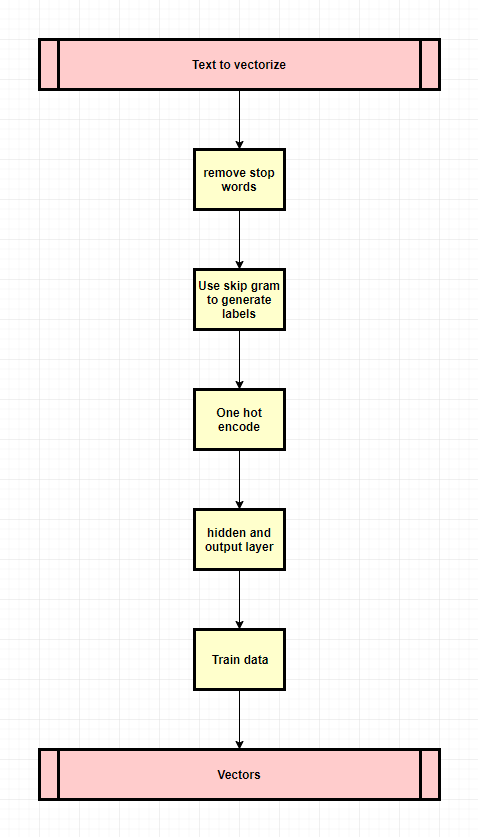
#### Word2vec

The word2vec (19) portion of the back end will run on a server as it is to intensive to run on the clients. Each new document will have to be trained separately as each document will produce a different set of vectors. Word2vec outputs word embeddings by predicting the neighbouring words given a word. Word2vec is a 3-layer neural network which consists of an input layer, a hidden layer and an output layer.



*Figure 13- Word2vec structure*

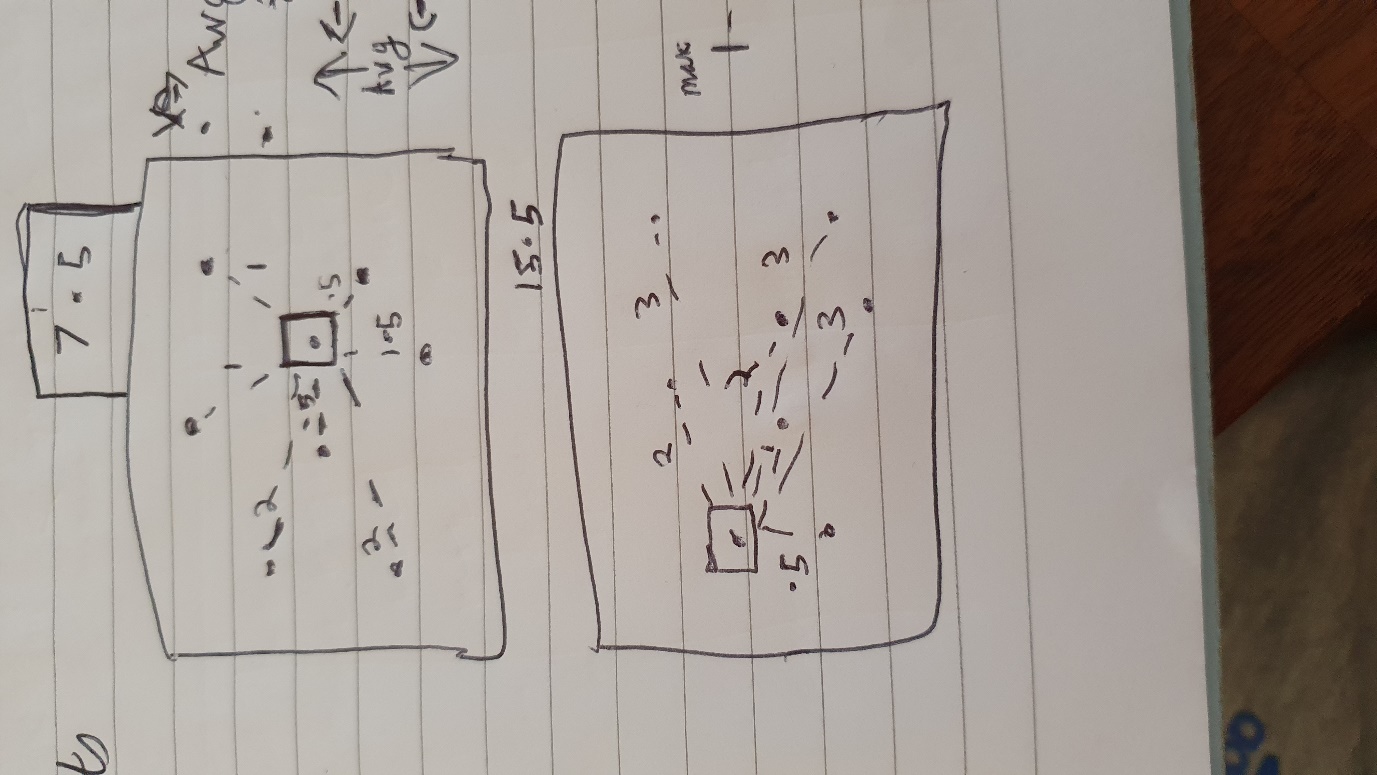
Training the model to predict the next word will never give high accuracy at the task. But the hidden layer which is the vectors used to predict the outcome give us the relationship of the words to each other.



*Figure 14 - word2vec Design*

#### Compare

This portion of the backend will be responsible for comparing the outputs of the image classification and word2vec. This logic will come in two steps, first the system will decide if the image could be an advertisement or not. This will be a Boolean value based of if the word is in the text. As words can be different but mean the same thing, querying the word net database is necessary. This will allow the system to match a greater variety of words.

The second stage of the compare logic is to decide how likely the possible advertisement is to be an actual advertisement. To do this the output of the word2vec algorithm is used. This is a two-dimensional graph of how each word in the text relates to each other. Using the Euclidean distance, we can compare how connected a word is to the document. Therefore, a more connected word will have a smaller total value for the lengths of its neighbouring points.

In the above diagram this idea is shown, the middle point is more connected meaning it will get a lower score whereas the second diagram the point is on the edge meaning the other points are further away making the value higher.

There are a few ways to calculate the distance between two points such as the Euclidean distance, Manhattan distance and Chebyshev Distance. Each of these methods have different quirks as they approach the problem differently. Because the Euclidean distance is the easiest to intuitively understand it was chosen as the method for calculating this metric, but this requires more mathematical operations so in future versions it may be quicker to use the Manhattan distance as this method requires less operations. The Euclidean distance can be equated to traveling along the hypotenuse of a right-angled triangle whereas the Manhattan distance can be equated to traveling along the other two sides.

## 3.7. Conclusions

In this chapter, the three main back end technologies were discussed was given for each of t, and an overall design hem. The middle tier will be responsible for connecting the backend to the frontend and translating the gathered information into meaningful data. The design of the UI was explored and fleshed out.

# 4. Experiment Development

## 4.1. Introduction

From my research and design, I have decided on a few technologies to use for this system. These technologies may change depending on how this experiment develops. Upon entering a new site scraping the DOM for the relevant data (the images and text of the site). This data is then shipped of to the flask API which is responsible for running the necessary backend functions

## 4.2. Software Development

The first step to developing any system is to set up version control. This project is being managed through GitHub. As the prototype of this system already has a Git repository, we will be extending from that.

## 4.3. Front-End

The front-end interface will be implemented in html, CSS and JavaScript. The main html file will be responsible for showing the user information about the system. Upon entering a new website, the chrome extension will run the script responsible for collecting and shipping the data to the middle tier Flask API. The returned information will then be displayed to the user. The chrome extension consists of a manifest file, the popup HTML and CSS and any scripts that should be run.

The manifest file is responsible for providing chrome with important information. The required information by chrome is the manifest version, name and version. The name is simply the name of the extension, the version is the version of the chrome extension and the manifest version is the version of the manifest file which as of Chrome 18 should be set to the integer 2.

## 4.4. Middle-Tier

Flask api

The middle tier of this system is a flask API. This system is responsible for accepting the POST and GET requests from the chrome extension and passing this data to the backend scripts that classify the data and decide if said data constitutes an advertisement or not. Each of these methods have a different route in the flask API. These routes along with the URL state the type of http traffic that the method will accept.

The POST methods of this system import the process management methods of the classifier scripts. When a POST request is made to the URL, Flask will run the method associated with that route which will call the necessary backend script. This data is then returned to flask in a dictionary format and sent back to the chrome extension.

The GET methods are responsible for explaining what the corresponding POST method does and the way in which the data should be structured for that method to be used.

<https://developer.mozilla.org/en-US/docs/Web/HTTP/CORS> As we are using Flask as a fetch API, CORS (Cross-Origin Resource Sharing) must be used for the chrome extension to communicate with said API. In order to do this, we must import the CORS flask library, set the CORS headers in our flask app to content type and before running the app convert into CORS.

## 4.5. Back-End

Word2vec, image classifier , wordnet.

#### Compare

Comparing the outputs of the afore mentioned algorithms is done using the wordnet database and by using Euclidean metrics.

## 4.6. Conclusions

# 5. Testing and Evaluation

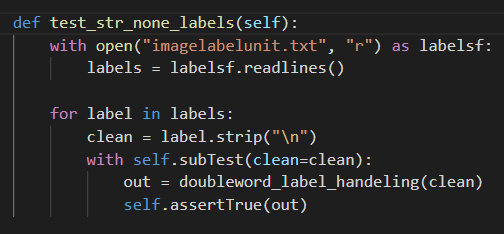
## 5.1. Introduction

In this chapter, the explanation of how unit testing occurred and how the final system was evaluated.

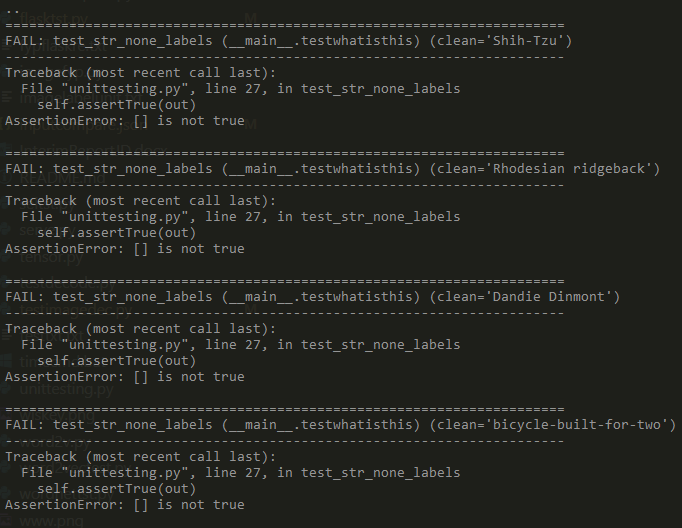
## 5.2. System Testing

In software development, testing is an important part of the development process. In the RAD methodology this is especially so. All software has bugs, to mitigate this testing after iterating is essential. Through out the development of this project both unit testing and ad hoc testing to validate the iterations of the programme.

Unit testing was most effective when applied to the code that decides what a word is. Using the labels of the image classification algorithm this function was tested and various changes were made because of this testing. For example, the original function did not consider double words. This was never a consideration as the word2vec model ensures a one-word format whereas in the image classifier 400 words are double words like red wine. The bug would not have been detected till much later or at all if not for the unit testing.



The above code is responsible for one of the subtests. The imagelabelunit.txt contains the labels of the image classifier, this is read in and cleaned up. For every word in this list, the wordnet database will be queried if there is no output then the query failed meaning there is a problem. The failure will be reported in the output of the unit test script.



The current state of the unit test is nine sub checks fail. Many of the words that fail are double barrel words or are irrelevant for the evaluation of this experiment as such this is unnecessary to solve this problem.

An ad hoc method of testing was required for the Flask API as with this application it either works or it doesn’t. To do this testing the curl tool was utilised as the client plugin had not been developed at the time. This ended up being a poor method of testing as the chrome plugin utilises CORS which is used when a web app must query a fetch API. <https://developer.mozilla.org/en-US/docs/Web/HTTP/CORS> The apparent successful testing of the API meant that when developing the POST functionally of the chrome extension the problems experienced seemed to purely be on the extension side of the application. After implementing the CORS package in the flask API and adding CORS mode, accept header and body json this issue was solved.

Another problem in testing the API in this manner is was in running the algorithms one at a time and then closing the process. This meant that when running the scripts in the flask API the process wold never close witch is the intended functionality. But the methods to free the resources are not guaranteed to function. These methods include TensorFlow’s build in methods like reset\_default\_graph() and close() but also using library’s like numba which have a Cuda implementation to select the GPU and close the current thread. As such, the best way to manage the memory consumption of these scripts is to create a new process and when they are finished processing the data end said process freeing the memory back to the system. This is guaranteed to happen as the process had ended. <http://numba.pydata.org/numba-doc/0.13/CUDADevice.html>

## 5.3. System Evaluation

5 chosen Websites and display results.

## 5.4. Conclusions

# 6. Conclusions and Future Work

## 6.1. Introduction

## 6.2. Conclusions

## 6.3. Future Work

#### Browser image classification

#### Retrain the inception model

The default model is the first runner up for the ImageNet Large Scale Visual Recognition Challenge. ImageNet is an image database organized according to the WordNet hierarchy. Retraining the model for more advertisement-based images would improve the accuracy for this application of the classifier. These advertisement-based images could be company logos. The logo may not be the main thing the image classifier identifies but if the prediction has a high enough percentage the system can be sure that the image is an advertisement. The image can then be removed without having to compare it with the text. This will make the system faster and more efficient as the main system will not have to be queried if the image classification is done in browser.

#### Lan implementation

The current implementation of this project is designed to run on a webserver and to be queried by a chrome plugin this idea came from current implementations of AdBlock in browsers. Other methods to solve this problem exist like Pi-hole as mentioned earlier and Privoxy which is a web proxy for modifying and filtering webpages. Following these projects ideas, A server which sits in-between the clients and the internet could modify pages achieving a similar solution.

The advantages of this approach would be the privacy of people using the system, as the system is no longer shipping the contents of the user’s webpages to an external server. If the server were to collect logs or was compromised this system could be used as another data point for the tracking and identification of the advertisement servers thus nullifying the purpose of the system this new implantation would mitigate this issue. Another advantage would be the mitigation of browser fingerprinting. This is the process of defining who you are by scraping the setup of your browser like the operating system you are running, the browser and its version, the language and time zone, plugins and many more. This new implementation would effectively make everyone on the network identical.

The biggest disadvantage of this approach is the current resources necessary to run the algorithms. The image classification portion of the system has been well optimised and is effective at classifying images but word2vec relies on training a new dataset on each query this takes considerable resources, making the required hardware expensive. As implementations of this algorithm get more efficient this implementation will become feasible to implement in home environments . Another problem with this approach is by default it won’t work with https traffic; this would have to be solved to make it an effective solution.

#### Removing advertisements

Now that the system can identify images that it thinks are advertisements the next logical step would be to remove them in some fashion. In its current iteration, this can be done with JavaScript by removing the element from the DOM. The implementation proposed above would remove the advertisements before it even gets to the client computer.

#### Word clouds

Word clouds are a simple way of visualising the frequency of a word in text. Early in the development of this project the frequency of the words was disregarded as a method to define what the text is about as in testing the top words tended to not describe the text was about and as such the system uses word2vec which takes every word in the text into account. But when it comes to text like blogs, tutorials, speeches, etc this method of text classification becomes much more viable. The method is also far more lightweight than word2vec meaning it could be implemented in browser either replacing word2vec on sites that contain blog in the URL or bolstering word2vec’s results on other sites.

# Bibliography