

OWL Predict AP2

Jude Dillon – B00737902
BEng Software Engineering
April 2021

Abstract

Overwatch league is an online esports league which hosts and streams between some of the top teams and players in the world and many people worldwide watch it. In 2020 “Overwatch League (OWL) Grand Finals drew 120,000 concurrent viewers on the esports competition’s live stream on YouTube.” [1] These viewers understand the general hierarchy between teams and as such like to predict the outcomes of games and even whole tournaments before they happen.

Sometimes though there are games with seemingly unpredictable outcomes where a team that before the game looked like they were almost certain to win may perform worse than expected or even lose. This may be due to differences in playstyle or strategy that allows a seemingly worse team to outperform certain teams with generally higher records because their playstyles counteract them.

Often when these outcomes occur viewers are shocked and surprised and wonder if their predictions could have been more accurate and how they could have expected it at all.

OWL Predict will take a different perspective on predicting game outcomes using the data of previous matches and predicting a winner purely based on the data without any human bias involved in the calculation. This could lead to more accurate predictions than even people well versed in the teams and the league could make.

People would be able to choose two teams and OWL Predict would tell them what it predicts to be the most likely outcome so they could compare that to their own predictions to enhance their decision.

The target audience for this product would be any Overwatch League viewers, especially those who like to discuss predictions and strategize how some teams may be able to improve their results with changes.

The aim of OWL Predict is to create a web-based application which will predict the outcome of an Overwatch League match between 2 teams using a Machine Learning algorithm based on data from previous games played in Overwatch League.

The system I have finalised with this project is a website that can make predictions of who will win an overwatch league game between two teams, in this current implementation it uses two predictors, and the user can change the season that predictions are based off, as well as the number of neighbours.

Acknowledgements

I would like to thank my mentor Dr Pat Corr for his guidance and knowledge throughout the project.

I would like to thank my peer support group for help with discussions and ideas.

I would like to thank my family for supporting me through this hard year.

Contents

Abstract	2
Acknowledgements	2
1 Requirement Control Document & Modification of the Project Plan	5
1.1 Final List of Requirements	5
1.1.1 Functional Requirements	5
1.1.2 Non-Functional Requirements	5
1.2 Requirements Evolution.....	6
1.3 Modifications Done to the Project Plan	6
1.3.1 Modified Gantt Chart	7
2 System Design	8
2.1 System Architecture Diagram	9
2.2 Interface Design	10
2.2.1 Wireframes	10
2.2.3 Consideration for Usability of the User Interface	11
2.3 Data Support Design	12
2.3.1 Consideration of Security and Data Validation.....	12
2.3.2 ER Diagram.....	13
2.4 User Interaction Design.....	14
3 System Implementation.....	15
3.1 Reflection on Implementation Plan	15
3.2 Tools and Languages Used	15
3.2.1 Python	15
3.2.2 Gitlab and Git Bash.....	15
3.2.3 VS Code	16
3.2.4 Mongo DB	16
3.2.5 PyMongo	16
3.2.6 Angular JS.....	16
3.2.7 Overwatch League Stats Lab Data	16
3.3 Evidence of Version Control.....	17
3.4 Volume of Code Produced	17
3.5 System Walkthrough.....	18
3.6 Consideration of Security Implementation	21
4 System Verification	22
4.1 Reflection on Verification Plan.....	22
4.2 Verification Results	22

4.3 Confirmation Statement of System Meeting Requirements	23
5 System Validation	23
5.1 Reflection of Validation Plan	23
5.2 Validation Results	23
5.3 Consideration for Future Work	23
6 Conclusion and Reflection.....	24
6.1 Project Appraisal	24
6.2 Reflection of Project Plan.....	24
6.3 Reflection of Initial Time/Effort Estimation.....	25
6.4 Reflection of Software Methodology	25
7 References	26
8 Appendices	27
8.1 Appendix A: Final Requirements Formal Format VOLERE	27
8.2 Appendix B: Additional Artefacts required for the Project.....	29
8.3 Appendix C: Code Document	30

1 Requirement Control Document & Modification of the Project Plan

1.1 Final List of Requirements

1.1.1 Functional Requirements

ID	Requirement	Priority	Risk Level
F1	User can select two different Overwatch League teams	Must Have	Low
F2	System will predict team that will win using a Machine Learning Algorithm	Must Have	High
F3	System will output team that it predicts to win	Must Have	Medium
F4	System needs to be able to extract data used for predictions from the dataset	Must Have	Medium
F5	System will provide accurate predictions (above 60% accuracy)	Should Have	Low
F6	System will output a percentage stating how sure it is of its prediction	Should Have	Low
F7	Users can tune the range of closest data points that the system uses to make decisions	Should Have	Low
F8	Users can choose which season of Overwatch League will be used to make predictions in the system	Should Have	Medium
F9	System will make prediction within 2 seconds [1]	Should Have	Medium
F10	User can make predictions through an API	Could Have	Low

Figure 1 Functional Requirements

1.1.2 Non-Functional Requirements

ID	Requirement	Priority	Risk Level
NF1	System will be robust	Must Have	Medium
NF2	System will be intuitive	Must Have	Low
NF3	System will look visually appealing	Must Have	Low
NF4	System must work on majority of browsers	Must Have	Medium

Figure 2 Non-Functional Requirements

1.2 Requirements Evolution

During the development process some changes were made to the initial requirements that were created during the initial project planning.

F8 was changed from “Users can choose which data will be used to make predictions in the system” to “Users can choose which season of Overwatch League will be used to make predictions in the system”. The justification for this change was that although the initial plan was for the user to be able to change both the seasons used for predictions and be able to choose different predictors for the system to use for predictions, the developer found that changing the predictors used for each prediction was much more difficult to implement within the time constraints than expected. As a result of this the project manager made the decision to instead implement the selection of which season the data used for predictions would be gathered from, so users would still be able to have a more specific prediction to their liking.

No other changes were made to the requirements because from discussions between the project manager and a focus group of target users the remaining requirements were all deemed sufficient for what members of the focus group expected from the product.

1.3 Modifications Done to the Project Plan

There were quite a number of modifications made to the project plan and these can be seen in the new Gantt chart in Figure 3. The developer decided to take a break over the Christmas season and instead of classifying more data for the use in KNN they decided to take the current initial implementation they had created at the end of AP1 and to convert it to work in the system before adding new functionality to it.

Due to unexpected illness in the developer’s family in January they were unable to maintain a level of focus to complete the earlier stages of the system in the planned time and although I still achieved them they took longer than originally expected. This included converting the initial implementation code to use Mongo DB; Creating the API which would access the database and make predictions; and the creation of the website frontend.

This hadn’t been accounted for in the risk assessment as it was seen as incredibly unlikely but it was a huge setback to progress on the project and led to changes to the requirements in order to meet time constraints of the whole project and meant that later sections seen to be lower priority were given less time for example making the frontend look visually appealing and not as many predictors as originally planned were able to be created for the KNN algorithm.

1.3.1 Modified Gantt Chart

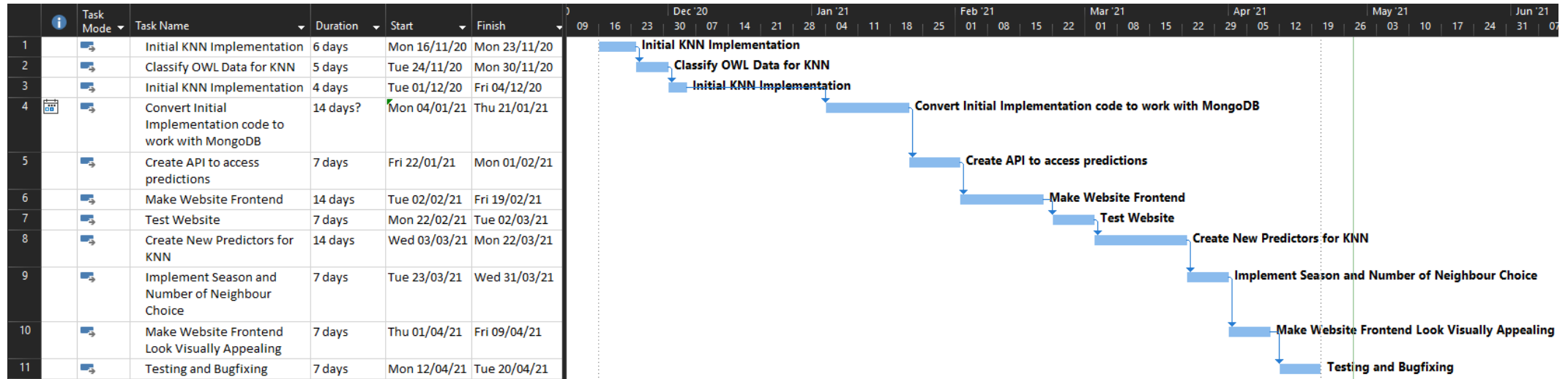


Figure 3 Modified Gantt Chart

2 System Design

The approach to the design of the architecture of the system was to make it all flow in a way where it moved the data from the frontend, to the backend via an API call which would access the database to then calculate the prediction with the K Nearest Neighbour algorithm [2] which would then return the prediction from the original API call to the frontend.

The overall plan for the design of the system from a user standpoint was to make it as simple and intuitive for the users as possible so it was easy for them to use, this was achieved with the use of abstraction [3].

The visuals, in particular the colours of the system were inspired by Overwatch League [4] and the colour scheme used in it because it would be familiar to users of the system.

2.1 System Architecture Diagram

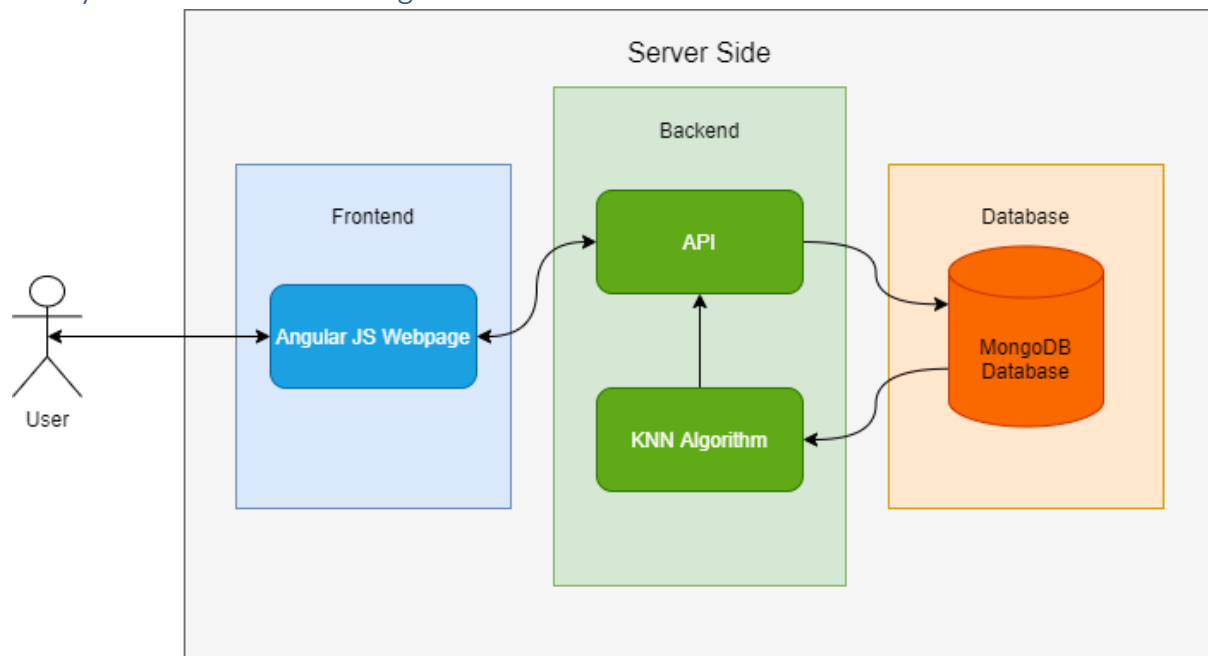


Figure 4 System Architecture Diagram

The system architecture diagram in Figure 4 shows an overall view of OWL Predict, how the data flows through the system and how each section of the system communicates with each of the other sections.

The user opens the website and the webpage which is created via angular JavaScript will display the values the user needs to input to make a prediction.

When the user inputs and submits the required values, the frontend will make a call containing these values to the API which is created in python.

The API will then convert these values to a format ready for the K Nearest Neighbour Algorithm and then send them to the K Nearest Neighbour Algorithm.

The K Nearest Neighbour Algorithm will send these values to the Mongo DB database to calculate the predictors for the inputted data as well as calculate the predictors for stored data when using the users selected season.

When these values are all returned from the database the K Nearest Neighbour Algorithm will create a list of all the K nearest neighbours to the input data where K is the number of neighbours selected and submitted by the user.

The algorithm will then gather the responses from this list of nearest neighbours and if the outcomes of this game were a win or a loss for team 1, if more of the nearest neighbours predict the outcome as a win then the algorithm will return to the API that it predicts a win for team 1 and its percentage confidence in it which is calculated as the percentage of nearest neighbours where the outcome was a win and vice versa for a win for team 2.

The API will then return the result of this prediction to the frontend where it will be displayed to the user.

2.2 Interface Design

2.2.1 Wireframes

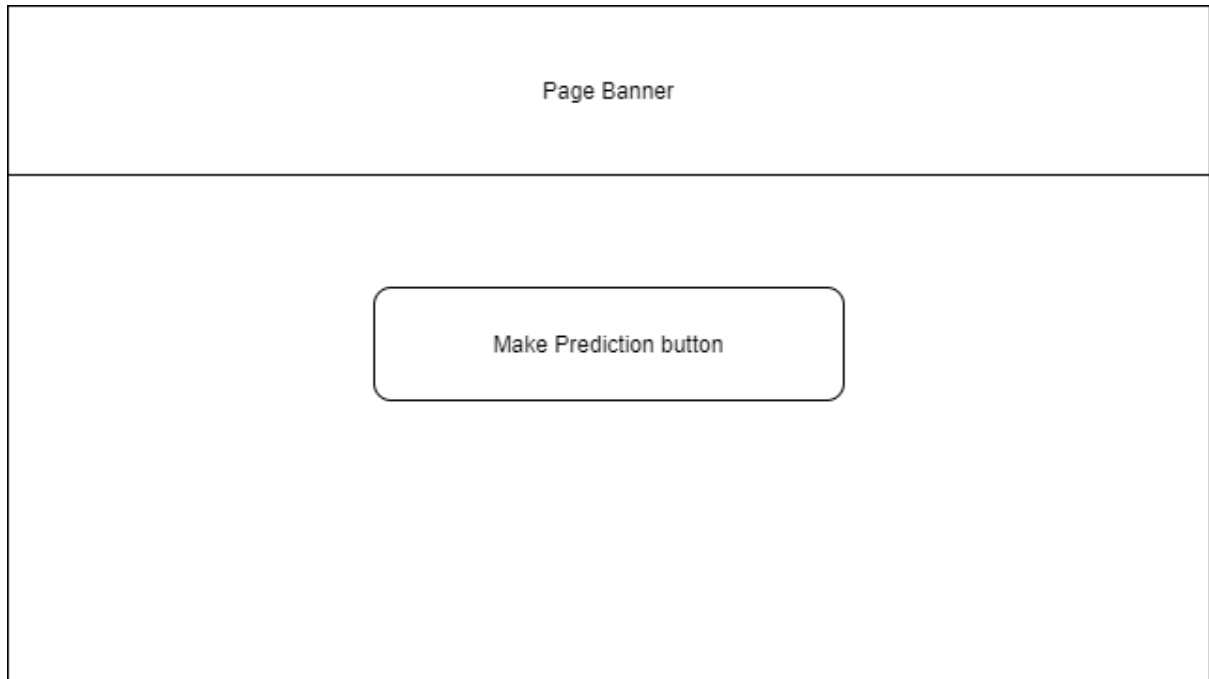


Figure 5 Home page wireframe

Figure 5 displays the home page. This page greets the user, displays the title of the system and if they would like to make a prediction, they can click the button which will then send them to the prediction page (Figure 6).

The wireframe illustrates the layout of the Prediction page. It features a 'Page Banner' at the top. Below it, a central form contains four stacked dropdown menus labeled 'Team 1 dropdown', 'Team 2 dropdown', 'Number of Neighbours dropdown', and 'Season dropdown'. A 'Submit button' is positioned below these menus. At the bottom of the form area is a 'Prediction Output' box.

Figure 6 Prediction page wireframe

Figure 6 displays the prediction page. This page allows the user to make a prediction by inputting their chosen values into the respective dropdown menus. The Team 1 and Team 2 dropdowns are mandatory and must not be left empty, if they are left empty or select the same teams or are not interacted with then the submit button will not appear and an error message will be displayed. When all values are valid and the submit button has been clicked then the system will make a prediction with the user's selected inputs and will then populate the prediction output which will be visible to the user.

2.2.3 Consideration for Usability of the User Interface

When designing and creating the user interface the project manager wanted to make the system as simple and intuitive as possible for users whilst also giving them as much control as possible over the way their predictions would be made so that both high- and low-level users could make use of the system.

To achieve this the system has few inputs to be made and the user simply has to select them from dropdown menus without any need to make an account to further streamline the use of the system. For low level users to be able to make predictions easily the fields for everything other than the teams is pre-filled out in the form which means that high level users can still tune these variables to their liking.

The developer also tried to make good use of colours similar to what the Overwatch league uses so that users would be familiar with the colours and understand the connection to Overwatch League.

2.3 Data Support Design

2.3.1 Consideration of Security and Data Validation

In this system the project manager and developer have taken care to not require logins or any personal data from users to use the system and as such there was no need to encrypt the database. It is also important to consider the possibility of DDoS attacks being able to take the system down but for the current implementation of the system the project manager is not putting protection in place for this as the website is not yet publicly hosted but in later implementations it will be important to take measures against this.

Due to the database running on Mongo DB there is a possibility of injection attacks to the system [5], the developer has taken measures to mitigate this risk through data validation. The only data the user can input through the frontend website is data from dropdown menus and the user cannot type their own inputs for these and the combination of data is validated on the page and can only be submitted if it is valid. In the current implementation there is no validation for predictions done through the API and this will be added in a later implementation.

2.3.2 ER Diagram

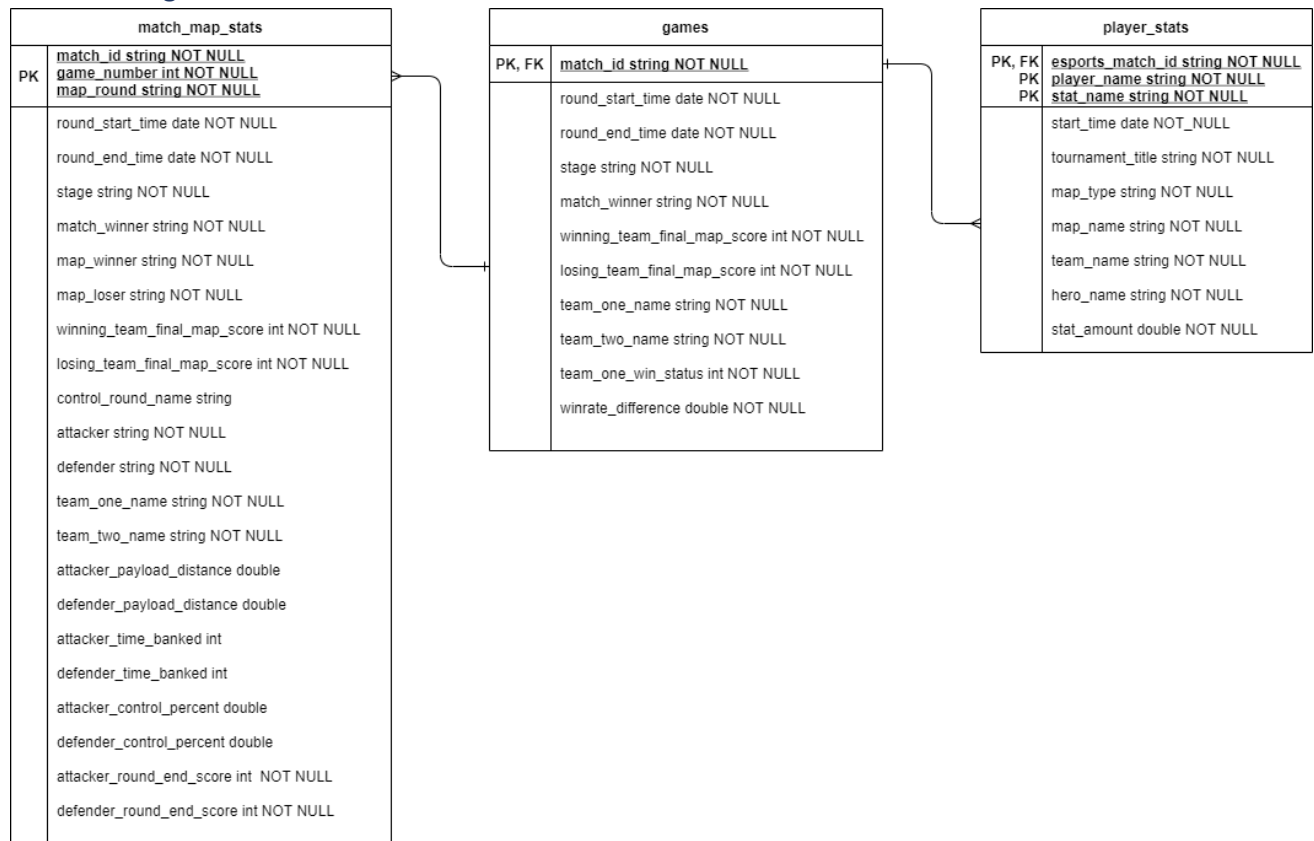


Figure 7 ER Diagram

Figure 7, shown above is the ER diagram for the database being used in OWL Predict. The match_map_stats table was created directly from the official Overwatch League Stats Lab data [6] where the developer downloaded their Map Stats dataset and imported it into Mongo DB, it contains an entity for each round of each map played in Overwatch League games.

The developer then created the games table from this dataset to have data on an individual game basis and each game contains multiple maps thus the many to one relationship between the match_map_stats table and the games table.

The developer originally created the dataset for this table with a python script adapted from the initial implementation in AP1 which would read in the match_map_stats.csv file and convert it into list format and it would then remove records with duplicate match_id's and create a new csv file from this. The developer then simply removed unnecessary properties for predictions manually using excel and imported the dataset into Mongo DB.

The player_stats table was like the match_map_stats table also created directly from the Overwatch League Stats Lab data where the developer downloaded each of their player stats datasets and imported them all into one Mongo DB table.

The table contains an entity for every individual player statistic for every player in each game of Overwatch League and as such there is a one to many relationship between the games table and the player_stats table.

2.4 User Interaction Design

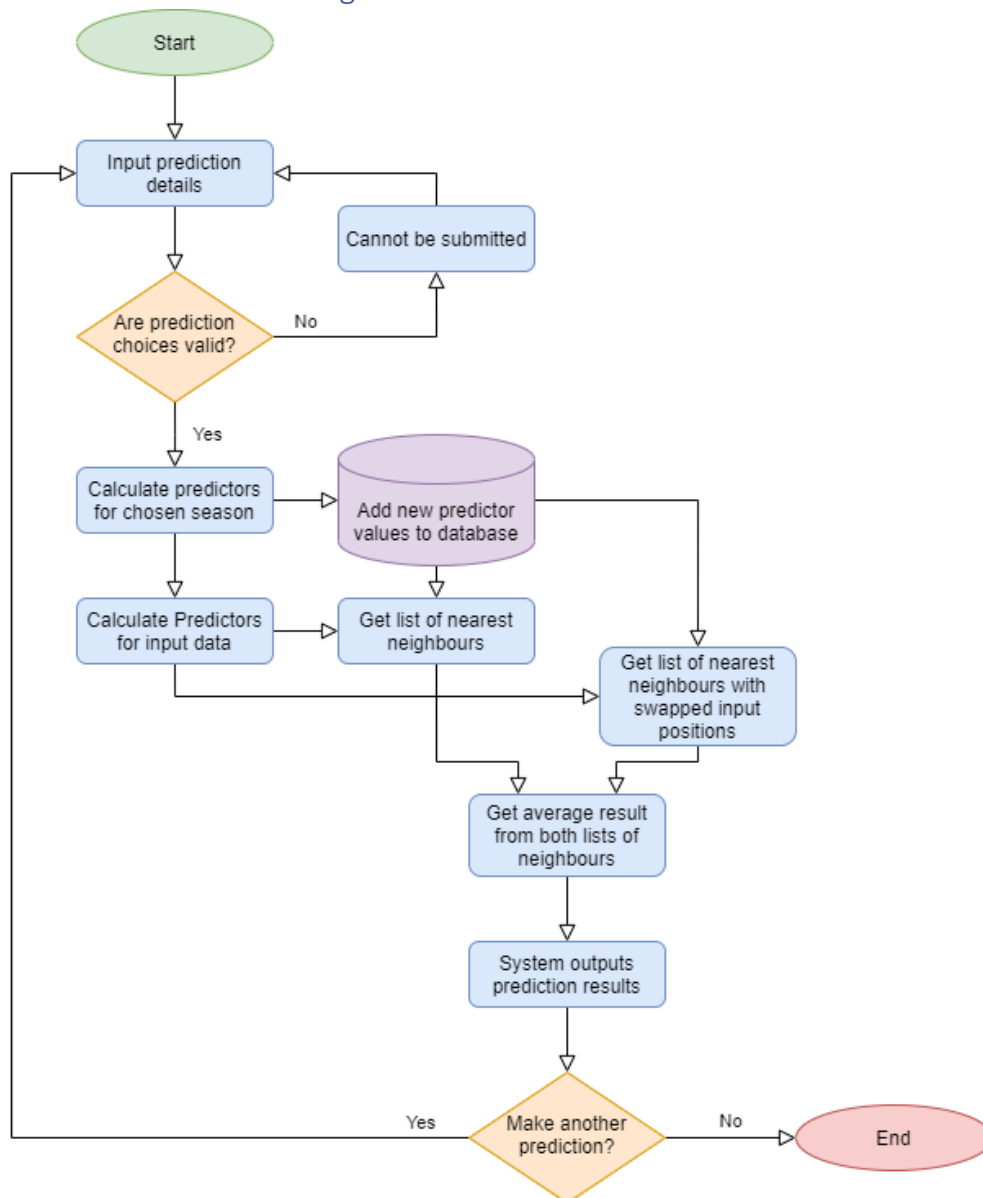


Figure 8 System flow diagram

Figure 8 shows the system flow diagram of the overall system making a prediction from a user's input.

When the user inputs their choices for the prediction they are checked for validity, if invalid the user will not be able to submit them. If they are valid then upon submission the predictors will be recalculated with the constraint of the user's chosen season and these new predictors will be stored in the database. These new predictors will then be used to create two lists of K nearest neighbours where K is selected from the user's input and the two lists will each have the order of the user's inputted teams swapped. The average results of both of these lists will then be used to predict which team is going to win as well as the percentage likelihood of them winning. This prediction will then be output to the user, at which point they can make another prediction if they would like to do so.

3 System Implementation

3.1 Reflection on Implementation Plan

Overall, with the implementation plan the project manager made the decision of using Kanban which served the project well as there were a number of changes to both the order of the steps in the implementation plan as well as the duration spent on certain steps in the plan and Kanban allowed for flexibility in both of these. One thing in particular that could have been planned better in the implementation plan would have been the choice of predictors as they were unexpectedly difficult to create and calculate especially when implementing them at such a late date in the code and perhaps taking care to make the earlier code more scalable would also have helped with this too.

3.2 Tools and Languages Used

3.2.1 Python

Python was used by the developer due to the developer's experience in using the language before. As well as their experience in using the language to create an API which was something the developer and project manager planned to have in their project from the beginning as a way to create the system. It was ideal in the creation of the system because it was very easy to manipulate the data in it through the use of dictionaries and the developer was able to create their own K nearest neighbour algorithm in python without the use of external libraries due to this. Although the developer was able to create this library themselves this also made things take more time than they may have otherwise and caused difficulties so it may have been more beneficial to use a python library for the K nearest neighbour algorithm, for example scikit learn [7] which may have made this easier but the developer may have lost some understanding of the algorithm if this had been used. Overall though the choice of python worked well for the development needs of the project and the developer was pleased with its performance even though it could have been better with the use of libraries.

3.2.2 Gitlab and Git Bash

Gitlab and git bash were used together for version control and maintaining a backup of the system code. The developer used gitlab as opposed to other version control software due to the quite large amount of free storage a user gets and because they had over a year of experience using it in work and other projects. The reason they used git bash was also due to familiarity and it meant they were able to very quickly move work from their computer to git without having to learn how to use it. Although the IDE used also had a way to access git via a GUI the developer was inexperienced in using a GUI for git and due to time constraints did not think it was necessary to learn it but perhaps if the developer had taken the time to learn it, they could have found it had a boost to their workflow speed rather than having to exit to a different program for git commits.

3.2.3 VS Code

VS Code was used by the developer as it is able to support a large number of different languages and as there were a number of different languages being using. This meant the developer did not have to switch to different IDE's to look at different parts of the system. VS Code is also quite lightweight and runs well on the developer's system compared to some other IDE's the developer has used. It has git integration so if the developer wanted to, they could access git from a GUI within VS Code but even without accessing the GUI it shows lines that have been changed from the last git version which helps to keep track of exactly what work has been done.

3.2.4 Mongo DB

Mongo DB was used by the developer for storing and making changes to the data which would be used for predictions. The reason for using it was that the developer had a large amount of experience in using it and how to integrate it into a python API and website. It was convenient to import csv or json files into Mongo DB which made importing the Overwatch League data much easier than other databases. One of the downsides of Mongo DB when compared to an SQL database, is that although the developer has more practical experience with Mongo DB, they had more experience with using SQL databases from university work and work experience so certain queries were quite hard to translate into a Mongo DB format.

3.2.5 PyMongo

PyMongo was used by the developer to integrate the python based API with Mongo DB. PyMongo is officially the recommended way to work with Mongo DB from python according to their own documentation [8] and the developer has used it before in projects.

3.2.6 Angular JS

Angular JS was used to help create the frontend for the system and to link it up with the API. The main reason for using Angular JS was that the developer also had experience in using it to make an API based website in the past and as such knew how to use it. Although there is a large potential for what can be created using Angular JS the developer had issues creating something more visually appealing within the time constraints as it was quite difficult to learn how to use some of the features of Angular JS.

3.2.7 Overwatch League Stats Lab Data

Overwatch League Stats Lab Data was used by the developer as data to make predictions from. This data is the official data gathered by the Overwatch League so there were no alternatives available for this. Although there were no alternatives to use the developer had issues with how the data was formatted which made it difficult to adapt for this system in certain circumstances and because it is only updated with post-match data it makes it impossible to have predictions that become more accurate as games are ongoing but again it was the only data available.

3.3 Evidence of Version Control

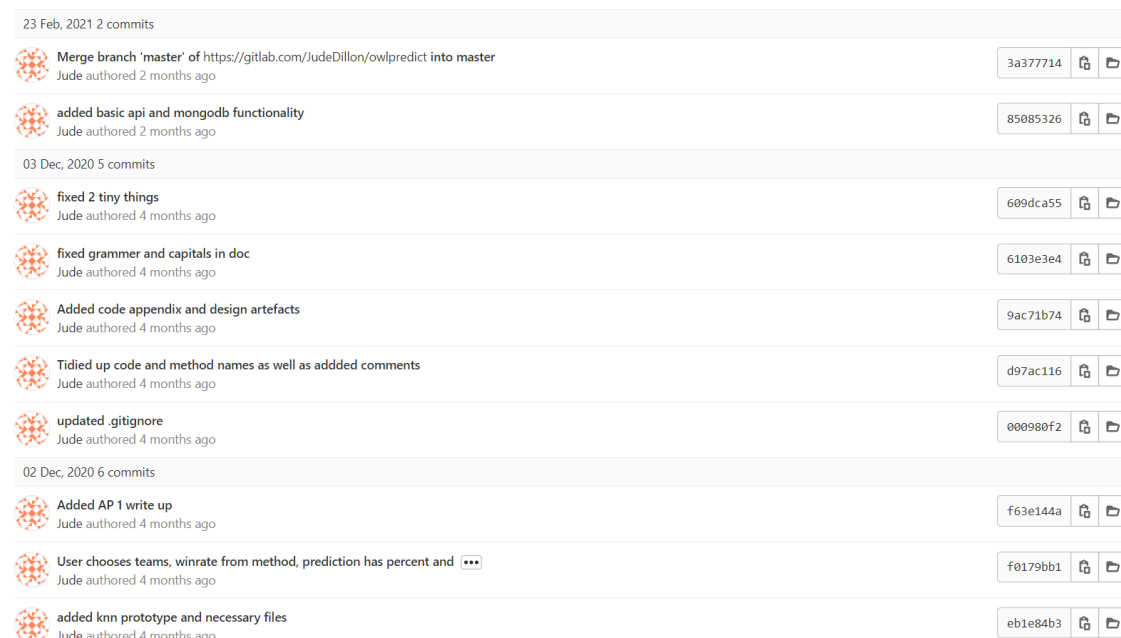


Figure 9 Git commit history

Figure 9 shows the git commit history. From the beginning of the initial implementation and throughout the development of the system the developer was using gitlab to maintain a backup of the system as well as for version control. This helped to mitigate any risk of hardware failure because if any system were to fail there would still be a recent backup available so that the whole system would not be lost. It also made it convenient for the developer to work on the system on multiple different systems which meant that they were able to move the work to a laptop and work elsewhere if necessary.

Although there was integration between git and VS code (the IDE the developer used) which would allow the developer to use a GUI to access git, the developer preferred to use git bash command line to make pushes and pulls to and from gitlab because he had more experience using this and it allowed his workflow to move smoothly.

As there was only one developer in this project there was no need to set up branch control but it would be useful to implement if in the future the project manager would like to add more developers to this project. It would ensure that all code added to the project was up to a high standard with the use of code reviews from peer developers.

Overall GitLab's version control was very useful to the project allowing the developer to rollback changes when needed and providing reassurance with the knowledge that if something were to go wrong there was an easily accessible way to either retrieve the code or undo a mistake.

3.4 Volume of Code Produced

Item	Amount
Python Methods	13
API Calls	2
Angular Components	6
Mongo DB Queries	9

Figure 10 Volume of Code Produced

3.5 System Walkthrough

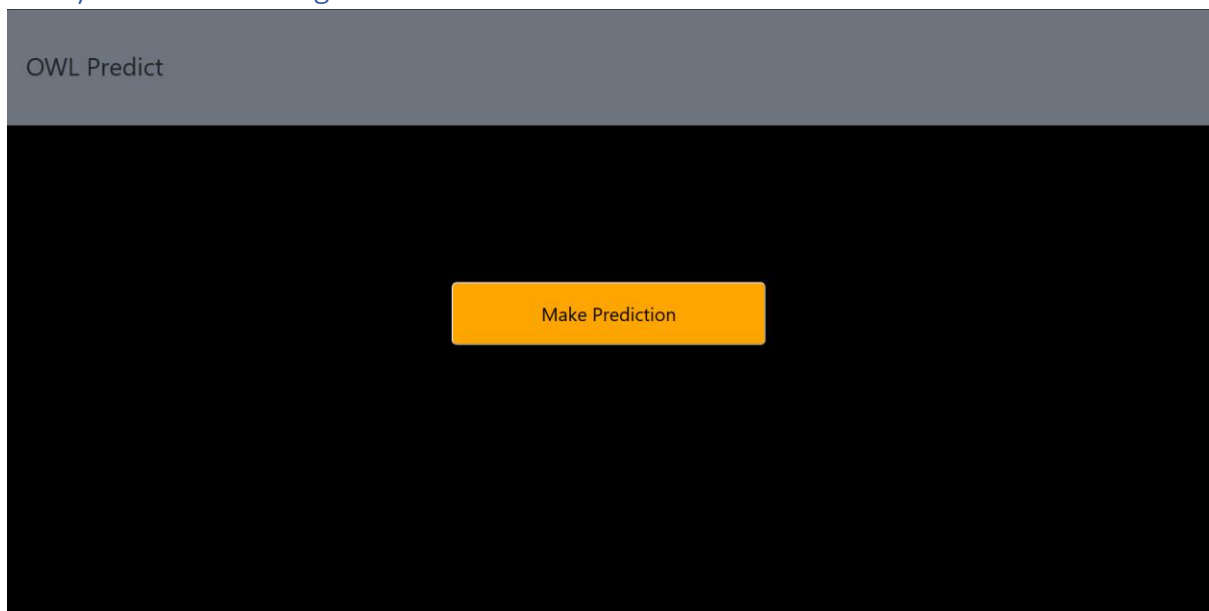


Figure 11 Home page

Figure 11 shows the home page of OWL Predict which will greet users when they open the website. If they click the Make Prediction button, they will open the prediction page.

The screenshot displays the prediction page of the 'OWL Predict' application. It has a dark grey header with 'OWL Predict' in white. The main content area is black and contains a form titled 'Please review this business' in white. The form includes four white input fields with dropdown arrows on the right: 'Team 1', 'Team 2', 'Number of neighbours' (which has '33' selected), and 'Season' (which has 'All Seasons' selected). Below these fields, a small white text label reads 'You must complete all fields'. At the bottom of the form, there is a solid orange horizontal bar.

Figure 12 Prediction page

Figure 12 shows the prediction page where the users can choose two teams, the number of neighbours for predictions and which season they want their prediction to be based off. By default, the number of neighbours selected is 33 and the season selected is all of them. If the user does not choose two different teams, they are unable to make a prediction.

The screenshot shows a web application titled "OWL Predict". Below the title, there is a heading "Please review this business". The form contains four dropdown menus: "Team 1" with "Dallas Fuel" selected, "Team 2" with "Shanghai Dragons" selected, "Number of neighbours" with "9" selected, and "Season" with "Season 3" selected. A blue "Submit" button is located below the "Season" dropdown. A thick orange horizontal bar is positioned at the bottom of the form area.

Figure 13 Completed prediction page

Figure 13 shows a completed prediction page. Once the user has input valid values the submit button will appear and when the user clicks submit a call will be made to the API to make a prediction.

```
@app.route("/predict/<string:team_one>/<string:team_two>", methods=["GET"])
def make_prediction(team_one, team_two):
    num_neighbours = 33
    season = ""

    if request.args.get('neighbours'):
        num_neighbours = int(request.args.get('neighbours'))
    if request.args.get('season'):
        season = str(request.args.get('season'))
    seasonrgx = re.compile('.*' + season + '.*')

    update_predictors(seasonrgx)

    winrate_difference = winrate_dif_calc(team_one, team_two, seasonrgx)

    team_one_average_final_score = calc_average_final_score(team_one, None, seasonrgx)
    team_two_average_final_score = calc_average_final_score(team_two, None, seasonrgx)
    average_final_score_difference = team_one_average_final_score - team_two_average_final_score

    inputvalue = [winrate_difference, average_final_score_difference]

    neighbours = get_neighbours(inputvalue, num_neighbours, seasonrgx)

    neighbors_votes = [neighbour["team_one_win_status"] for neighbour in neighbours]

    team_one_win_votes = 0
    team_two_win_votes = 0
    for vote in neighbors_votes:
        if(vote == 1):
            team_one_win_votes+=1
        else:
            team_two_win_votes+=1

    if(team_one_win_votes > team_two_win_votes):
        prediction = "I predict that " + team_one + " wins by " + str(team_one_win_votes/len(neighbors_votes)*100) + "% of the neighbours votes"
    else:
        prediction = "I predict that " + team_two + " wins by " + str(team_two_win_votes/len(neighbors_votes)*100) + "% of the neighbours votes"

    return make_response(jsonify(prediction), 200)
```

Figure 14 Prediction API

Figure 14 shows the prediction API, when called it would first update all the predictors in the database to match the season chosen by the user. It would then create a list of the nearest neighbouring datapoints to this input and make a prediction based off the average outcomes of these games.

```

def update_predictors(season):
    winrate_list = []
    total_average_damage_done_list = []
    average_final_score_difference_list = []

    for game in games.find({"stage": season}):
        team_one = game["team_one_name"]
        team_two = game["team_two_name"]
        round_start_time = game["round_start_time"]
        match_id = game["match_id"]

        team_one_winrate = calc_winrate(team_one, round_start_time, season)
        team_two_winrate = calc_winrate(team_two, round_start_time, season)
        winrate_diff = team_one_winrate - team_two_winrate
        winrate_list.append(winrate_diff)

        team_one_average_final_score = calc_average_final_score(team_one, round_start_time, season)
        team_two_average_final_score = calc_average_final_score(team_two, round_start_time, season)
        average_final_score_diff = team_one_average_final_score - team_two_average_final_score
        average_final_score_difference_list.append(average_final_score_diff)

    minmax = list()
    for i in range(len(average_final_score_difference_list)):
        value_min = min(average_final_score_difference_list)
        value_max = max(average_final_score_difference_list)
        minmax.append([value_min, value_max])

    for i in range(len(average_final_score_difference_list)):
        average_final_score_difference_list[i] = (average_final_score_difference_list[i] - minmax[i][0]) / (minmax[i][1] - minmax[i][0])

    i = 0
    for game in games.find({"stage": season}):
        games.update_one(
            {"match_id": game["match_id"]},
            {"$set": {
                "winrate_difference": winrate_list[i],
                "average_final_score_difference": average_final_score_difference_list[i]
            }}
        )
        i = i + 1

```

Figure 15 update_predictors method

Figure 15 shows the update predictors method, when called it would calculate the average winrate difference for all the chosen games as well as the average final score difference and then would update these values in the database.

```

def get_neighbours(inputvalue, num_neighbours, season):
    distances = list()
    for game in games.find({"stage": season}):
        distance = get_distance(inputvalue, game)
        distances.append((game, distance))

    distances.sort(key=lambda tup: tup[1])

    neighbours = list()
    for i in range(num_neighbours):
        neighbours.append(distances[i][0])
    return neighbours

```

Figure 16 get_neighbours method

Figure 16 shows the get neighbours method which would find all the games in the dataset and sort them by their distance to the input data.

```
def get_distance(inputvalue, game):
    distance = 0.0
    row2 = [(game["winrate_difference"]), (game["average_final_score_difference"])]
    for i in range(len(inputvalue)-1):
        distance += (inputvalue[i] - row2[i])**2
    return sqrt(distance)
```

Figure 17 get_distance method

Figure 17 is the get_distance method, it calculates the Euclidean displacement of the stored predictors from the input data and then squares and gets the square root of it to remove any negative values, thus converting it into distance.

The screenshot shows a web application titled "OWL Predict". Below the title, there is a section titled "Please review this business" with four dropdown menus: "Team 1" (Dallas Fuel), "Team 2" (Shanghai Dragons), "Number of neighbours" (5), and "Season" (Season 3). A blue "Submit" button is located below the dropdowns. Below the button, a yellow box displays the prediction result: "I predict that Shanghai Dragons wins by 88.8888888888889% of the neighbours votes".

Figure 18 Prediction page with prediction

Figure 18 shows the prediction page after the prediction has been completed and the results are then output to the screen.

3.6 Consideration of Security Implementation

Due to no personal data or user logins/passwords being stored the database does not need to be encrypted to comply with GDPR regulations and there is a lowered risk of data breaches. Although the current implementation of the system does not have any protection against Mongo DB injection attacks the developer would need to consider adding protection for this in future implementations to ensure that the data is not altered by an external party.

4 System Verification

4.1 Reflection on Verification Plan

The verification plan for the most part involved testing each new part of the system as it was implemented and then at the end of the development of the whole project. The developer spent some time testing edge cases of the system.

Overall this verification plan worked well and suited the developer's development process but due to unforeseen circumstances in earlier stages of the system's development there was not as much time as expected to test edge cases.

The main issues the project manager can see with this method of verification is that the developer may miss some test cases that another perspective might be able to find. Due to time constraints this method of verification was seen as the best option by the project manager.

4.2 Verification Results

Test#	Test Description	Req. tested	Expected result	Actual result	Pass/Fail
1	Make a prediction with no teams selected	F1	Cannot submit choices	Cannot submit choices	Pass
2	Make a prediction with only one team selected	F1	Cannot submit choices	Cannot submit choices	Pass
3	Make a prediction with two of the same teams selected	F1	Cannot submit choices	Cannot submit choices	Pass
4	Make a prediction with two different teams selected	F1, F3, F6	Successful prediction shows predicted winner and percentage	Successful prediction shows predicted winner and percentage	Pass
5	Make 2 predictions with the same teams change the season selected	F4, F8	Changing the season should change the prediction result	Changing the season changes the prediction result	Pass
6	Make 2 predictions with the same teams change the number of neighbours selected	F7	Changing the number of neighbours should change the prediction result	Changing the number of neighbours should change the prediction result	Pass
7	Make prediction and time how long it takes to receive the result	F9	Should take less than 2 seconds	Time is sometimes more than 2 seconds and sometimes less	Fail
8	Make a prediction through the API	F10	Using the API to make a prediction is successful	Using the API to make a prediction is successful	Pass

4.3 Confirmation Statement of System Meeting Requirements

After completing the verification, the project manager has decided that the system sufficiently meets its requirements. The requirements that the system has failed to meet are F9 because the system has to do more calculations than expected and the project manager underestimated how many would be needed. The developer was also unable to complete verification of F5 because it was difficult to confirm the accuracy of the system within the time constraints.

5 System Validation

5.1 Reflection of Validation Plan

Originally the validation was planned to be done via black box testing from a focus group of target users after each new feature was added and then interviewing them. After further consideration from the project manager they instead decided to just do one round of black box testing and interviewing at the finalised state of the system for this implementation. The project manager saw this as a more valuable way to get the opinions of the target users as well as a faster way to complete the validation and it was useful for the validation process.

The developer had planned to create a test script to test the accuracy of the predictions being made by the system but due to time constraints was unable to create this script to validate the accuracy levels for the current implementation of the system.

5.2 Validation Results

From the black box testing with the focus group of users who all used the system at the same time the project manager gathered from them that the system was robust. This is because there were no issues with it withstanding the load of multiple users at once and that the system worked on the majority of browsers as the focus group were using a number of different browsers.

From the interviews with the focus group of users afterwards the project manager found that all the users reported the system as being very intuitive and simple to use and nobody had any difficulties using the system. Although the focus group of users reported that the visual design of the webpage was not as appealing as they would have liked. Some users were also disappointed that there was no accuracy reported for the predictions of the system as without those they felt unable to trust the result.

5.3 Consideration for Future Work

Some considerations to be made for future work on this system would be that for later implementations to create a test script to test the accuracy of predictions to show validity of the system's predictions to users. Which may entice them into being more likely to use it and to take further consideration and time with the visual design of the webpage which would require the developer to spend more time learning how to use Angular JS.

6 Conclusion and Reflection

6.1 Project Appraisal

Overall, I am quite proud of what I have achieved with this project. The aim of this project from the beginning was to create a web-based application which will predict the outcome of an Overwatch League match between 2 teams using a Machine Learning algorithm based on data from previous games played in Overwatch League. Although this system is not complete, I feel that I have with this release got the system in a form where it fulfils its most fundamental requirements which the initial aim set out to do.

I was able to use knowledge I had accrued from other subjects throughout university and from my placement job. Not only did I use this knowledge and experience, but I also built upon it further with creating the backend and frontend of the project as well as when managing and maintaining the project and the version control of it.

I learned a lot about machine learning and AI throughout the creation and research of this system. Before beginning this project I was intimidated by machine learning but I now feel like I have a good grasp on a number of different machine learning algorithms and would like to in future work more with them.

During the creation of this system Overwatch League announced an official partnership with IBM where they were using AI with IBM's Watson to give all players and teams in the Overwatch League a ranking according to their AI analysis [9]. I feel this is very similar to my project and this further validates my initial thoughts that there was demand for a system like this. In comparison to my system which only uses 2 different predictors in its current implementation IBM's system uses over 360 predictors and prioritises them too. I believe the way the user is able to tune mine and compete teams against each other means that it still has advantages of its own and given more time could possibly scale to a similar size.

I was a bit disappointed in my project that I was unable to meet the requirement of how long it takes the system to make a prediction and with how it looked visually but I have plans in my head for how I could in future releases improve both of these things with optimisations to my K nearest neighbour algorithm. I would also like to create a model to measure the accuracy of the predictions made by the system as I had originally planned.

6.2 Reflection of Project Plan

The project plan worked quite well for both the planning of the AP1 write up and then both the development and write up of AP2. It had to undergo a number of changes due to unexpected circumstances but thanks to how I had structured the plan originally it was easy to make these changes to it.

Before I began this project although I had done some informal planning before I had never formally planned out my work to anything close to this degree which was a big learning curve for me and I learned a large amount about planning a project and now feel I would be much better equipped to do so again in the future.

I was disappointed that I was unable to stick to the original project plan that I had made in AP1 but I feel like from persevering through and restructuring my plan to better suit the time I had available I feel like I learned some very valuable lessons that I would not have otherwise learned.

6.3 Reflection of Initial Time/Effort Estimation

I think that my initial time and effort estimations in this project were where I went most wrong. I vastly underestimated how difficult implementing and learning certain new things would be, but even so I gained valuable experience in how to better estimate my own capabilities and I think that in future I will continue to gradually get better at making time or effort estimations as I become both a more experienced developer and gain more experience in completing projects.

Due to my misgivings with how much time or effort would be required for certain parts of the project it led to me regretfully being unable to complete certain parts to the standard I had originally envisaged but thanks to the risk assessment and the requirements I completed I was able to ensure I focused on the top priorities for the project.

6.4 Reflection of Software Methodology

The methodology I chose of Kanban was extremely well suited to the project and to my workflow. I think of all the choices in this project planning this is the one I was most proud of. I had a whiteboard in my room beside the computer with the tasks that needed to be completed on it and as I did them I would tick them off the whiteboard. This helped me to stay motivated as I could see the progress I was making with my work and how much more there was left to do. It also allowed with more flexibility with the project plan and how I allocated my time to each task because I was able to dynamically change it depending on how much effort or time a task was requiring which was especially useful with tasks I needed time to research the tools for.

Although the software methodology of Kanban worked well for me for this project I think in future if I was to do a project with a team I may instead opt for a methodology like AGILE or something else instead because I think with more people it may be hard to keep track of who is completing what task in Kanban.

The only weakness I could see with Kanban in this project for me would be that it could be possible to get carried away and spend too much time on a singular task leaving many tasks to the end so it was important that when I decided I needed to spend more time on a specific task that I reworked the entire project plan to fit around that as best I could.

7 References

- [1] F. F.-H. Nah, A Study on Tolerable Waiting Time: How Long Are Web Users Willing to Wait?, Association for Information Systems, 2003, p. 285.
- [2] N. S. Altman, "An Introduction to Kernel and Nearest-Neighbor Nonparametric Regression," *The American Statistician*, vol. 46, no. 3, pp. 175-185, 1992.
- [3] Sharpened Productions, "Abstraction Definition," TechTerms, 19 April 2019. [Online]. Available: <https://techterms.com/definition/abstraction>. [Accessed 26 04 2021].
- [4] Wikipedia, "Overwatch League," Wikipedia, 26 04 2021. [Online]. Available: https://en.wikipedia.org/wiki/Overwatch_League. [Accessed 26 04 2021].
- [5] Z. Banach, "What is NoSQL Injection and How Can You Prevent It?," NetSparker, 29 May 2020. [Online]. Available: <https://www.netsparker.com/blog/web-security/what-is-nosql-injection/>. [Accessed 29 04 2021].
- [6] Overwatch League, "Overwatch League Stats Lab: Beta," 26 04 21. [Online]. Available: <https://overwatchleague.com/en-us/statslab>. [Accessed 27 04 21].
- [7] scikit-learn, "scikit-learn," scikit-learn, [Online]. Available: <https://scikit-learn.org/stable/>. [Accessed 27 04 21].
- [8] MongoDB, "MongoDB Python Drivers," MongoDB, [Online]. Available: <https://docs.mongodb.com/drivers/python/>. [Accessed 28 04 28].
- [9] IBM, "IBM and the Overwatch League," IBM, [Online]. Available: <https://www.ibm.com/sports/overwatchleague/>. [Accessed 29 04 2021].

8 Appendices

8.1 Appendix A: Final Requirements Formal Format VOLERE

Requirement: 1	Requirement Type: Functional
Description: User can select two different Overwatch League teams	
Rationale: There needs to be 2 teams selected to make a prediction of the winner	
Source: Developer – Jude Dillon	
Fit Criterion: System will fulfil if user can select 2 different teams to make a prediction	
Customer Satisfaction Rating: 5	Customer Dissatisfaction Rating: 0
History: October 2020	

a

Requirement: 2	Requirement Type: Functional
Description: System will predict team that will win using a Machine Learning Algorithm	
Rationale: The aim of the project is to make the predictions using Machine Learning	
Source: Developer – Jude Dillon	
Fit Criterion: System will fulfil if it uses a Machine Learning Algorithm to make predictions	
Customer Satisfaction Rating: 5	Customer Dissatisfaction Rating: 0
History: October 2020	

Requirement: 3	Requirement Type: Functional
Description: System will output team that it predicts to win	
Rationale: The user needs to be able to see the outcome of the prediction	
Source: Developer – Jude Dillon	
Fit Criterion: System will fulfil if outcome of prediction is displayed to the user	
Customer Satisfaction Rating: 5	Customer Dissatisfaction Rating: 0
History: October 2020	

Requirement: 4	Requirement Type: Functional
Description: System needs to be able to extract data used for predictions from the dataset	
Rationale: If the system cannot calculate the prediction data itself it will need to be inserted manually each time it makes a prediction using new values	
Source: Developer – Jude Dillon	
Fit Criterion: System will make new calculations for changing the season without requiring new data to be input	
Customer Satisfaction Rating: 5	Customer Dissatisfaction Rating: 0
History: October 2020	

Requirement: 5	Requirement Type: Functional
Description: System will provide accurate predictions (above 60% accuracy)	
Rationale: It would be good to boast a high accuracy of predictions for users to see	
Source: Developer – Jude Dillon	
Fit Criterion: System will fulfil if prediction accuracy can be calculated and is above 60%	
Customer Satisfaction Rating: 1	Customer Dissatisfaction Rating: 4
History: October 2020	

Requirement: 6	Requirement Type: Functional
Description: System will output a percentage stating how sure it is of its prediction	
Rationale: This will help the user gauge how likely a prediction is to be accurate	
Source: Developer – Jude Dillon	
Fit Criterion: System will fulfil if predictions are displayed to the user with a percentage of their confidence alongside the prediction	
Customer Satisfaction Rating: 4	Customer Dissatisfaction Rating: 1
History: October 2020	

Requirement: 7	Requirement Type: Functional
Description: Users can tune the range of closest data points that the system uses to make decisions	
Rationale: There needs to be a choice of how many nearest data points are used so the user can tune the predictions to their liking	
Source: Developer – Jude Dillon	
Fit Criterion: System will fulfil if user can choose number of neighbours when submitting a prediction	
Customer Satisfaction Rating: 5	Customer Dissatisfaction Rating: 0
History: October 2020	

Requirement: 8	Requirement Type: Functional
Description: Users can choose which season of Overwatch League will be used to make predictions in the system	
Rationale: There needs to be a choice of which seasons are used so the user can tune the predictions to their liking	
Source: Developer – Jude Dillon	
Fit Criterion: System will fulfil if user can choose season when submitting a prediction	
Customer Satisfaction Rating: 5	Customer Dissatisfaction Rating: 0
History: February 2021	

Requirement: 9	Requirement Type: Functional
Description: System will make prediction within 2 seconds	
Rationale: I want the website to feel responsive	
Source: Developer – Jude Dillon	
Fit Criterion: System will fulfil if predictions are made within 2 seconds	
Customer Satisfaction Rating: 2	Customer Dissatisfaction Rating: 3
History: October 2020	

Requirement: 10	Requirement Type: Functional
Description: User can make predictions through an API	
Rationale: Could allow other projects to integrate with my system	
Source: Developer – Jude Dillon	
Fit Criterion: System will fulfil if user can make predictions through an API	
Customer Satisfaction Rating: 3	Customer Dissatisfaction Rating: 2
History: October 2020	

Requirement: 11	Requirement Type: Non-Functional
Description: System will be robust	
Rationale: The system needs to be able to support a reasonable number of users at once	
Source: Developer – Jude Dillon	
Fit Criterion: System will fulfil if it can support a reasonable number of users at once	
Customer Satisfaction Rating: 4	Customer Dissatisfaction Rating: 1
History: October 2020	

Requirement: 12	Requirement Type: Non-Functional
Description: System will be intuitive	
Rationale: The system needs to be intuitive so a new user can use it without requiring help	
Source: Developer – Jude Dillon	
Fit Criterion: System will fulfil if users find the system easy to use	
Customer Satisfaction Rating: 5	Customer Dissatisfaction Rating: 0
History: October 2020	

Requirement: 13	Requirement Type: Non-Functional
Description: System will look visually appealing	
Rationale: The system needs to look nice to entice users to use it	
Source: Developer – Jude Dillon	
Fit Criterion: System will fulfil if user think it looks visually appealing	
Customer Satisfaction Rating: 1	Customer Dissatisfaction Rating: 4
History: October 2020	

Requirement: 14	Requirement Type: Non-Functional
Description: System must work on majority of browsers	
Rationale: I want users to be able to access the system regardless of browser	
Source: Developer – Jude Dillon	
Fit Criterion: System will fulfil if it works on most modern browsers	
Customer Satisfaction Rating: 4	Customer Dissatisfaction Rating: 1
History: October 2020	

8.2 Appendix B: Additional Artefacts required for the Project

8.3 Appendix C: Code Document

owlpredict\app.py

```
from flask import Flask, request, jsonify, make_response
from pymongo import MongoClient
from flask_cors import CORS
from bson import ObjectId
from math import sqrt
import re

app = Flask(__name__)
CORS(app)

client = MongoClient("mongodb://127.0.0.1:27017")
db = client.OWLPredict      #select the database
games = db.games

@app.route("/", methods=["GET"])
def index():
    return make_response( jsonify("Hello world"), 200)

def winrate_dif_calc(team_one, team_two, season):
    team_one_winrate = calc_winrate(team_one, None, season)
    team_two_winrate = calc_winrate(team_two, None, season)

    winrate_dif = team_one_winrate - team_two_winrate

    return winrate_dif

def calc_winrate(team, round_start_time, season):
    total_games = 0
    total_wins = 0

    if(round_start_time is None):
        total_wins = games.find({"$and":[{"match_winner":team}, {"stage": season}]}).count()
        total_games = games.find({"$and":[{"$or":[{"team_one_name":team}, {"team_two_name":team}]}], {"stage": season}}).count()
    else:
        for game in games.find({"$and":[{"$and":[{"$or":[{"team_one_name":team}, {"team_two_name":team}]}], {"stage": season}], {"round_start_time":{"$lt":round_start_time}}}):
            total_games+=1
            if((game["team_one_win_status"]==1 and game["team_one_name"]==team) or (game["team_one_win_status"]==0 and game["team_two_name"]==team)):
```

```

        total_wins+=1

    try:
        winrate = total_wins / total_games
        #if total games is 0 then winrate is 0
    except ZeroDivisionError:
        winrate = 0

    return winrate

def get_distance(inputvalue, game):
    distance = 0.0
    row2 = [(game["winrate_difference"]), (game["average_final_score_difference"])]
    for i in range(len(inputvalue)-1):
        distance += (inputvalue[i] - row2[i])**2
    return sqrt(distance)

def get_neighbours(inputvalue, num_neighbours, season):
    distances = list()
    for game in games.find({"stage": season}):
        distance = get_distance(inputvalue, game)
        distances.append((game, distance))

    distances.sort(key=lambda tup: tup[1])

    neighbours = list()
    for i in range(num_neighbours):
        neighbours.append(distances[i][0])
    return neighbours

def calc_average_final_score(team, round_start_time, season):
    total_games = 0
    total_score = 0

    if(round_start_time is None):
        for game in games.find({"$and":[{"$or":[{"team_one_name":team}, {"team_two_name":team}]}], {"stage": season}}):
            if((game["team_one_win_status"]==1 and game["team_one_name"]==team) or (game["team_one_win_status"]==0 and game["team_two_name"]==team)):
                total_score = total_score + game["winning_team_final_map_score"]
            else:
                total_score = total_score + game["losing_team_final_map_score"]
    ]
    total_games = games.find({"$and":[{"$or":[{"team_one_name":team}, {"team_two_name":team}]}], {"stage": season}}).count()
    else:

```

```

        for game in games.find({"$and":[
            {"$and":[
                {"$or":[
                    {"team_one_name":team},
                    {"team_two_name":team}]]},
                {"stage": season}]]},
            {"round_start_time":{"$lt":round_start_time}})):
            total_games += 1
            if((game["team_one_win_status"]==1 and game["team_one_name"]
            ==team) or (game["team_one_win_status"]==0 and game["team_two_name"]==team))
            :
                total_score = total_score + game["winning_team_final_m
ap_score"]
            else:
                total_score = total_score + game["losing_team_final_ma
p_score"]

        try:
            average_score = total_score / total_games
            #if total_games is 0 then winrate is 0
        except ZeroDivisionError:
            average_score = 0

        return average_score

def update_predictors(season):
    winrate_list = []
    total_average_damage_done_list = []
    average_final_score_difference_list = []

    for game in games.find({"stage": season}):
        team_one = game["team_one_name"]
        team_two = game["team_two_name"]
        round_start_time = game["round_start_time"]
        match_id = game["match_id"]

        team_one_winrate = calc_winrate(team_one, round_start_time, season)
        team_two_winrate = calc_winrate(team_two, round_start_time, season)
        winrate_diff = team_one_winrate - team_two_winrate
        winrate_list.append(winrate_diff)

        team_one_average_final_score = calc_average_final_score(team_one, roun
d_start_time, season)
        team_two_average_final_score = calc_average_final_score(team_two, roun
d_start_time, season)
        average_final_score_diff = team_one_average_final_score - team_two_ave
rage_final_score
        average_final_score_difference_list.append(average_final_score_diff)

```



```

minmax = list()
for i in range(len(average_final_score_difference_list)):
    value_min = min(average_final_score_difference_list)
    value_max = max(average_final_score_difference_list)
    minmax.append([value_min, value_max])

for i in range(len(average_final_score_difference_list)):
    average_final_score_difference_list[i] = (average_final_score_difference_list[i] - minmax[i][0]) / (minmax[i][1] - minmax[i][0])

i = 0
for game in games.find({"stage": season}):
    games.update_one(
        {"match_id": game["match_id"]},
        {"$set" :{
            "winrate_difference": winrate_list[i],
            "average_final_score_difference": average_final_score_difference_list[i]
        }}
    )
    i = i + 1

@app.route("/predict/<string:team_one>/<string:team_two>", methods=["GET"])
def make_prediction(team_one, team_two):
    num_neighbours = 33
    season = ""

    if request.args.get('neighbours'):
        num_neighbours = int(request.args.get('neighbours'))
    if request.args.get('season'):
        season = str(request.args.get('season'))
    seasonrgx = re.compile('.*' + season + '.*')

    update_predictors(seasonrgx)

    winrate_difference = winrate_dif_calc(team_one, team_two, seasonrgx)

    team_one_average_final_score = calc_average_final_score(team_one, None, seasonrgx)
    team_two_average_final_score = calc_average_final_score(team_two, None, seasonrgx)
    average_final_score_difference = team_one_average_final_score - team_two_average_final_score

    inputvalue = [winrate_difference, average_final_score_difference]

    neighbours = get_neighbours(inputvalue, num_neighbours, seasonrgx)

```

```

        neighbors_votes = [neighbour["team_one_win_status"] for neighbour in neighbors]

        team_one_win_votes = 0
        team_two_win_votes = 0
        for vote in neighbors_votes:
+         if(vote == 1):
            team_one_win_votes+=1
        else:
            team_two_win_votes+=1

        if(team_one_win_votes > team_two_win_votes):
            prediction = "I predict that " + team_one + " wins by " + str(team_one_win_votes/len(neighbors_votes)*100) + "% of the neighbours votes"
        else:
            prediction = "I predict that " + team_two + " wins by " + str(team_two_win_votes/len(neighbors_votes)*100) + "% of the neighbours votes"

        return make_response(jsonify(prediction), 200)

if __name__ == "__main__":
    app.run(debug=True)

```

owlpredict\frontend\src\app\app.component.ts

```

import { Component } from '@angular/core';

@Component({
  selector: 'app-root',
  templateUrl: './app.component.html',
  styleUrls: ['./app.component.css']
})
export class AppComponent {
  title = 'OWL Predict';
}

```

owlpredict\frontend\src\app\app.module.ts

```
import { NgModule } from '@angular/core';
import { BrowserModule } from '@angular/platform-browser';
import { ReactiveFormsModule } from '@angular/forms';

import { AppComponent } from './app.component';
import { PredictComponent } from './predict.component';
import { WebService } from './web.service';
import { HttpClientModule } from '@angular/common/http';
import { RouterModule } from '@angular/router';
import { HomeComponent } from './home.component';

var routes = [
  {
    path: '',
    component: HomeComponent
  },
  {
    path: 'prediction',
    component: PredictComponent
  }
];

@NgModule({
  declarations: [
    AppComponent,
    PredictComponent,
    HomeComponent
  ],
  imports: [
    BrowserModule,
    HttpClientModule,
    RouterModule.forRoot(routes),
    ReactiveFormsModule
  ],
  providers: [WebService],
  bootstrap: [AppComponent]
})
export class AppModule { }
```

owlpredict\frontend\src\app\home.component.css

```
.my-button{
  width: 500px;
  height: 100px;
  background-color: orange;
  border-radius: 8px;
}

.center {
  margin: 0;
  position: absolute;
  top: 50%;
  left: 50%;
  -ms-transform: translate(-50%, -50%);
  transform: translate(-50%, -50%);
}
```

owlpredict\frontend\src\app\home.component.html

```
<div class="jumbotron bg-secondary">
  <h1>OWL Predict</h1>
</div>

<div class="center">
  <button class="my-button" style="font-size: 30px;">Make Prediction</button>
</div>
```

owlpredict\frontend\src\app\home.component.ts

```
import { Component } from '@angular/core';

@Component({
  selector: 'home',
  templateUrl: './home.component.html',
  styleUrls: ['./home.component.css']
})
export class HomeComponent {
  ngOnInit() {
    document.body.style.backgroundColor = 'black';
  }
}
```

owlpredict\frontend\src\app\predict.component.css

```
.error { background-color: #fff0f0; }

.my-card{background-color: orange;}

.my-container{background-color: black;}
```

owlpredict\frontend\src\app\predict.component.html

```
<div class="jumbotron bg-secondary">
  <h1>OWL Predict</h1>
</div>

<div class="container">
  <div class="row">
    <div class="col-sm-12 text-white">
      <h2>Please review this business</h2>

      <form [formGroup]="predictForm" (ngSubmit)="onSubmit()">
        <div class="form-group">
          <label for="team1">Team 1</label>
          <select id="team1" name="team1" class="form-
control" formControlName="team1" [ngClass]="{'error': isValid('team1')}">
            <option value="Atlanta Reign">Atlanta Reign</option>
            <option value="Boston Uprising">Boston Uprising</option>
n>
            <option value="Chengdu Hunters">Chengdu Hunters</option>
n>
            <option value="Dallas Fuel">Dallas Fuel</option>
            <option value="Florida Mayhem">Florida Mayhem</option>
            <option value="Guangzhou Charge">Guangzhou Charge</option>
ion>
            <option value="Hangzhou Spark">Hangzhou Spark</option>
            <option value="Houston Outlaws">Houston Outlaws</option>
n>
            <option value="London Spitfire">London Spitfire</option>
n>
            <option value="Los Angeles Gladiators">Los Angeles Gla
diators</option>
            <option value="Los Angeles Valiant">Los Angeles Valian
t</option>
            <option value="New York Excelsior">New York Excelsior<
/option>
            <option value="Paris Eternal">Paris Eternal</option>
            <option value="Philadelphia Fusion">Philadelphia Fusio
n</option>
```

```

        <option value="San Francisco Shock">San Francisco Shock</option>
        <option value="Seoul Dynasty">Seoul Dynasty</option>
        <option value="Shanghai Dragons">Shanghai Dragons</option>
        <option value="Toronto Defiant">Toronto Defiant</option>
        <option value="Vancouver Titans">Vancouver Titans</option>
        <option value="Washington Justice">Washington Justice</option>
    </select>
</div>
<div class="form-group">
    <label for="team2">Team 2</label>
    <select id="team2" name="team2" class="form-control" formControlName="team2" [ngClass]="{'error': isValid('team2')}">
        <option value="Atlanta Reign">Atlanta Reign</option>
        <option value="Boston Uprising">Boston Uprising</option>
        <option value="Chengdu Hunters">Chengdu Hunters</option>
        <option value="Dallas Fuel">Dallas Fuel</option>
        <option value="Florida Mayhem">Florida Mayhem</option>
        <option value="Guangzhou Charge">Guangzhou Charge</option>
        <option value="Hangzhou Spark">Hangzhou Spark</option>
        <option value="Houston Outlaws">Houston Outlaws</option>
        <option value="London Spitfire">London Spitfire</option>
        <option value="Los Angeles Gladiators">Los Angeles Gladiators</option>
        <option value="Los Angeles Valiant">Los Angeles Valiant</option>
        <option value="New York Excelsior">New York Excelsior</option>
        <option value="Paris Eternal">Paris Eternal</option>
        <option value="Philadelphia Fusion">Philadelphia Fusion</option>
        <option value="San Francisco Shock">San Francisco Shock</option>
        <option value="Seoul Dynasty">Seoul Dynasty</option>
        <option value="Shanghai Dragons">Shanghai Dragons</option>
        <option value="Toronto Defiant">Toronto Defiant</option>
    </select>

```

```

        <option value="Vancouver Titans">Vancouver Titans</option>
        <option value="Washington Justice">Washington Justice</option>
    </select>
</div>
<div class="form-group">
    <label for="numberOfNeighbours">Number of neighbours</label>
    <select id="numberOfNeighbours" name="my-dropdown" class="form-control" formControlName="numberOfNeighbours">
        <option *ngFor="let number of numbers" [value]="number">{{number}}</option>
    </select>
</div>
<div class="form-group">
    <label for="season">Season</label>
    <select id="season" name="season" class="form-control" formControlName="season">
        <option value="">All Seasons</option>
        <option value="poo">Season 1</option>
        <option value="poo">Season 2</option>
        <option value="OWL 2020 Regular Season">Season 3</option>
    </select>
</div>
<span *ngIf="isIncomplete()">You must complete all fields</span>
<button *ngIf="!isIncomplete()" type="submit" class="btn btn-primary">Submit</button>
</form>
</div>
</div>
</div>
</div>
<div class="container" style="margin-top:20px;">
    <div class="row">
        <div class="col-sm-12">
            <div class="my-card text-black mb-3">
                <h1 class="card-header">
                    {{webService.prediction | async}}
                </h1>
            </div>
        </div>
    </div>
</div>
</div>

```

owlpredict\frontend\src\app\predict.component.ts

```
import { Component } from '@angular/core';
import { WebService } from '../web.service';
import { FormBuilder, Validators } from '@angular/forms';

@Component( {
  selector: 'predict',
  templateUrl: './predict.component.html',
  styleUrls: ['./predict.component.css']
})
export class PredictComponent {

  predictForm;
  numbers = [];

  constructor(public webService: WebService, private formBuilder: FormBuilder) {
    this.numbers = Array(50).fill(0).map((x,i)=>i+1);
  }

  ngOnInit() {
    document.body.style.backgroundColor = 'black';
    this.predictForm = this.formBuilder.group({
      team1: ['', Validators.required],
      team2: ['', Validators.required],
      numberOfNeighbours: 33,
      season: ''
    });
    this.webService.getHelloWorld();
  }

  onSubmit()
  {
    this.webService.getPrediction(this.predictForm.value.team1, this.predictForm.value.team2, this.predictForm.value.numberOfNeighbours, this.predictForm.value.season);
  }

  isValid(control) {
    return this.predictForm.controls[control].invalid && this.predictForm.controls[control].touched;
  }

  isUntouched() {
    return this.predictForm.controls.team1.pristine || this.predictForm.controls.team2.pristine;
  }
}
```



```

    isIncomplete() {
        return this.isInvalid('team1') || this.isInvalid('team2') || this.isUn
touched();
    }
}

```

owlpredict\frontend\src\app\web.service.ts

```

import { HttpClient } from '@angular/common/http';
import { Injectable } from '@angular/core';
import { Subject } from 'rxjs';

@Injectable()
export class WebService {
    private privateHello;
    private helloSubject = new Subject();
    hello = this.helloSubject.asObservable();

    private privatePrediction;
    private predictionSubject = new Subject();
    prediction = this.predictionSubject.asObservable();

    constructor(private http: HttpClient) {}
    getHelloWorld() {
        return this.http.get('http://localhost:5000/').subscribe(response=>
        {
            this.privateHello = response;
            this.helloSubject.next(this.privateHello);
        })
    }

    getPrediction(team1, team2, numberOfNeighbours, season) {
        return this.http.get('http://localhost:5000/predict/' + team1 + '/' + te
am2 + '?neighbours=' + numberOfNeighbours + '&season=' + season).subscribe(res
ponse=>
        {
            this.privatePrediction = response;
            this.predictionSubject.next(this.privatePrediction);
        })
    }
}

```