

# D&D Character Predictor

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## Introduction

In this report, a **D&D character dataset** was used to create a **D&D character predictor** that can be used to predict a character's class from its ability scores.

The **D&D character dataset** consists of approximately 2600 D&D characters obtained from GitHub (<https://github.com/oganm/dnddata>) with the permission of the repository creator.

The raw data was then filtered, wrangled, visualized, and two models were trained. A final **D&D character predictor** with an **accuracy** of **~50%** was obtained.

## Basic Introduction to Dungeons and Dragons

Dungeons and Dragons is a very popular role-playing game. Each player creates a character that has, among other attributes, a class, a race, and several ability scores. The twelve main classes in D&D are: barbarian, bard, cleric, druid, fighter, monk, paladin, ranger, rogue, sorcerer, wizard, and warlock. The six ability scores are: strength (str), dexterity (dex), constitution (con), intelligence (int), wisdom (wis), and charisma (cha). Ability scores are generally between 1 (very low) and 20 (very high), but there are some exceptions. Ability scores are chosen/assigned when the character is created and can be increased as a character levels up.

A player will usually assign his or her character's ability scores based on the class that they have selected for that character. For example, a fighter's highest ability score is usually strength, while a wizard's highest ability score is usually intelligence. Therefore, it should be possible to predict a character's class by examining its ability scores.

To simplify the dataset, the characters were limited to the main races in D&D, which are: dwarf, elf, halfling, human, dragonborn, gnome, half-elf, half-orc, and tiefling.

## Data Analysis and Model Development

### Create the Dataset

The raw dataset *dnd\_chars\_singleclass* was pulled directly from the GitHub repository. The raw dataset was then filtered down into the *character\_data* by selecting the class, race, str, dex, con, int, wis, and cha columns.

### Character Data

##	class	race	str	dex	con	int	wis	cha
## 1	Wizard	Human	8	12	20	20	14	10
## 2	Warlock	Kenku	12	20	16	14	14	21
## 3	Fighter	Orc	20	15	20	6	12	10
## 4	Paladin	Orc	20	13	20	10	12	14
## 5	Paladin	Dragonborn	20	8	16	8	8	20
## 6	Barbarian	Aarakocra	103	101	103	99	100	99

Looking at the *character\_data* dataset, we can see that it contains more classes and races than just the main ones listed above. We can also see that characters with classes or races other than the main ones occur less often.

### Character Classes

```
## # A tibble: 19 x 2
##   class      n
##   <chr>    <int>
## 1 Artificer      9
## 2 Barbarian    233
## 3 Bard         192
## 4 Blood Hunter    6
## 5 Cleric       272
## 6 commoner        1
## 7 Crafting Commoner 1
## 8 Druid        175
## 9 Fighter       361
## 10 Gunslinger     1
## 11 Monk          180
## 12 Mystic          8
## 13 Paladin       228
## 14 Ranger        196
## 15 Revised Ranger  13
## 16 Rogue         297
## 17 Sorcerer      168
## 18 Warlock       153
## 19 Wizard        216
```

### Character Races

```
## # A tibble: 19 x 2
##   race      n
##   <chr>    <int>
## 1 ""         3
## 2 1/2 Elf--Var. 1
## 3 Aarakocra    22
## 4 Aasimar       7
## 5 Air Genasi    3
## 6 Animal Hybrid  3
## 7 Animal Hybrid (Bear) 1
## 8 Beast         1
## 9 Bugbear      16
## 10 Caprien (Goatfolk) 1
## 11 Centaur      13
## 12 Changeling    8
## 13 Changeling 2.0 1
## 14 Child Of The Old Ones 1
## 15 "Copper Dragon " 1
## 16 Dark Elf      52
## 17 Deep Gnome    21
## 18 Dhampir       1
## 19 Dhampire       1
```

To facilitate the model, characters with unusual or uncommon races and classes were removed from the data set. Races in D&D can have sub-races, for example, a character can simply be a dwarf, or a hill dwarf, or a mountain dwarf, etc. Because of this, the dataset has been limited to the main races and their respective sub-races, though race is not used as a predictor for the model.

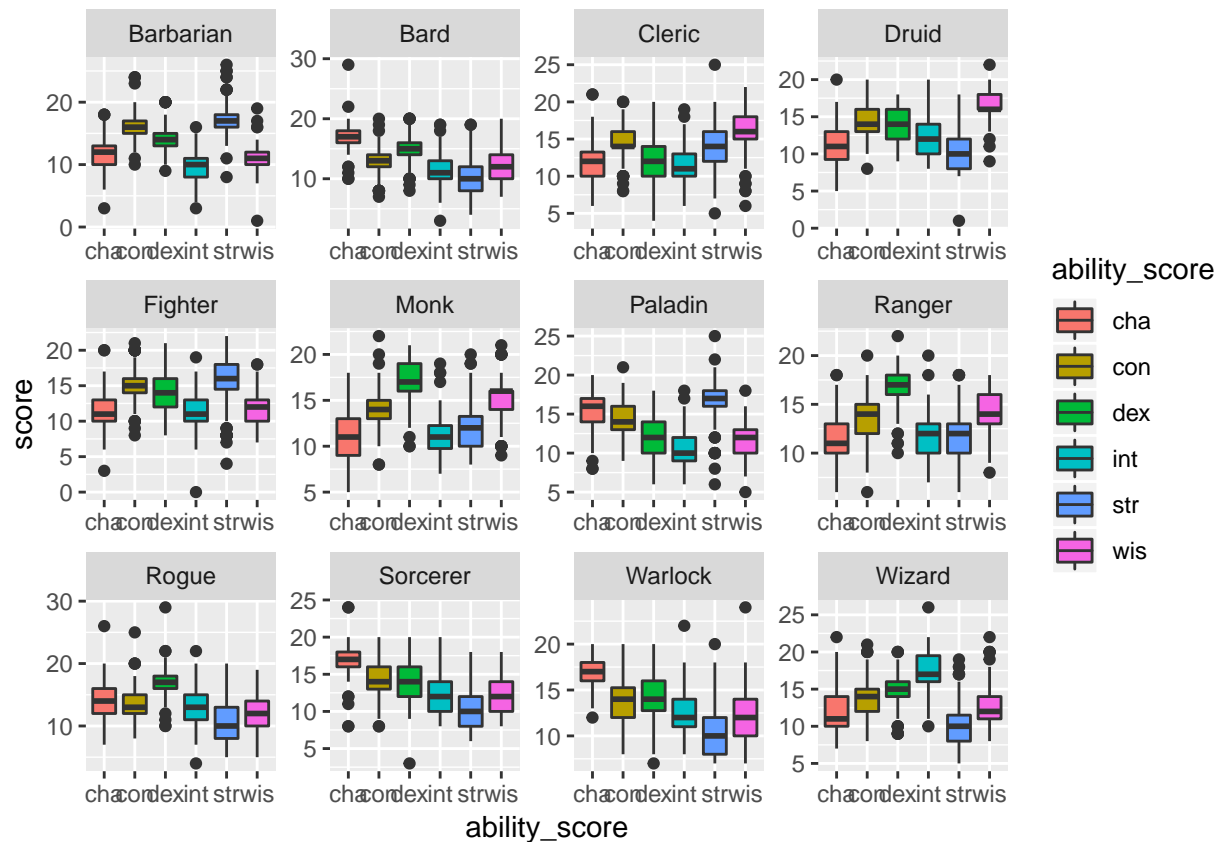
## Character Data

##	class	race	str	dex	con	int	wis	cha
## 1	Wizard	Human	8	12	20	20	14	10
## 2	Paladin	Dragonborn	20	8	16	8	8	20
## 3	Druid	Human	8	18	18	12	20	8
## 4	Wizard	Human	13	19	19	21	15	20
## 5	Wizard	Human	14	17	21	26	22	22
## 6	Wizard	Human	12	13	14	22	20	20

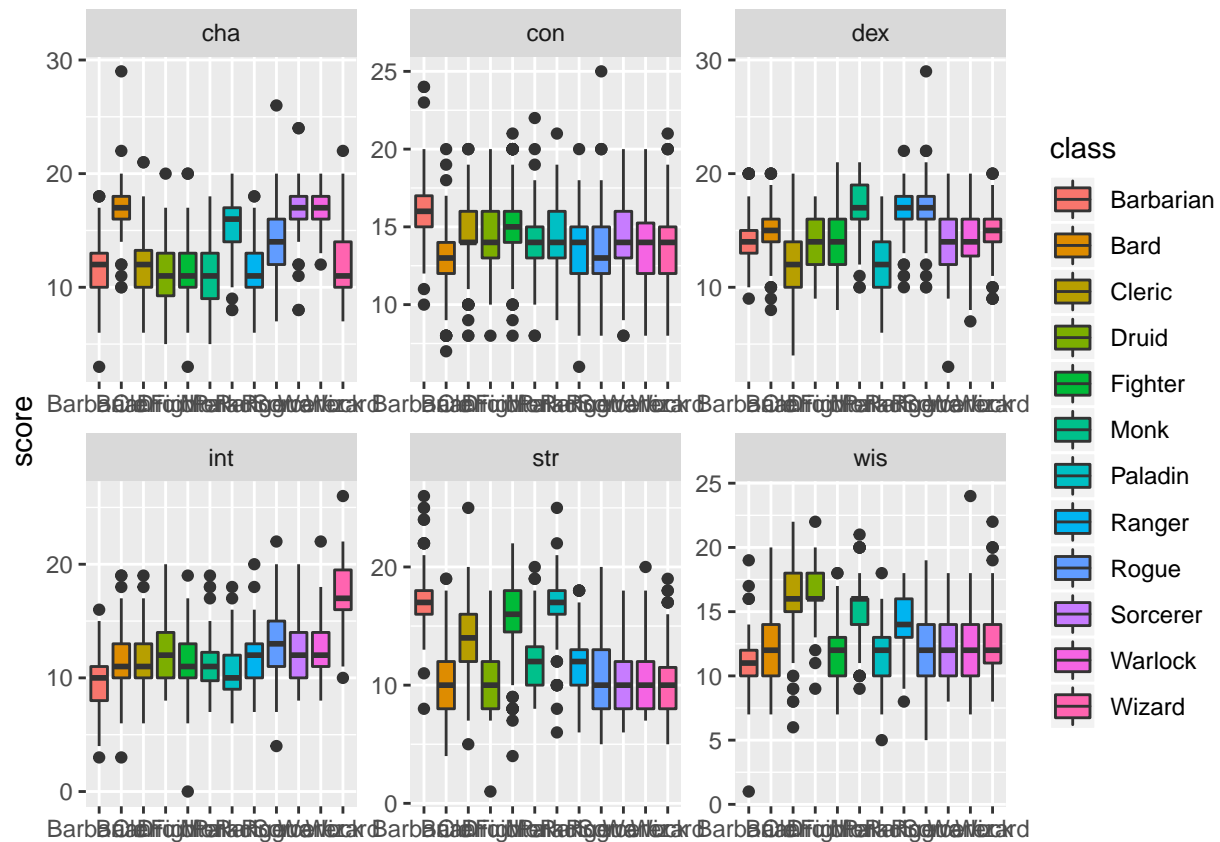
## Cursory Data Visualizations and Analysis

Plotting the dataset by class shows that, as expected, different classes tend to favor different ability scores. Barbarians generally have high strength and low intelligence, rogues typically have high dexterity and low strength, and wizards tend to have low strength and high intelligence.

### Ability Scores by Class



## Class by Ability Scores



## A First Model

The first model tested was the k-nearest neighbors (kNN) method tuned for k.

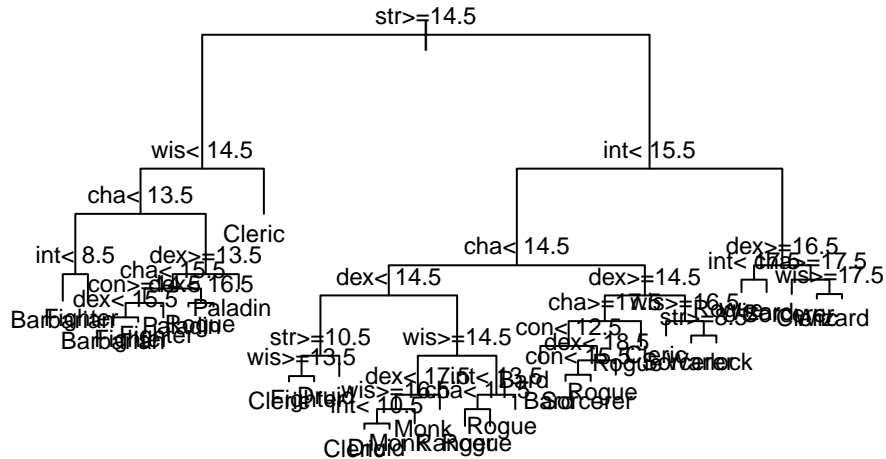
### kNN Model

The accuracy of the **tuned kNN** model is 0.49.

## A Second Model

The second model tested was a cp-optimized regression tree.

### Optimized Regression Tree



The accuracy of the **optimized regression tree** model is 0.45, which is a loss in accuracy from the previous model. Also, this regression tree is very crowded, almost impossible to read, and has way too many nodes. The excess of nodes seems to stem from the fact that even within a class, the values of the ability scores vary greatly. To accommodate this, the ability score data was ranked.

## Ranking the Ability Scores

A user defined function was created using the `rank()` function to rank a character's ability scores from highest (6) to lowest (1). The `apply()` function was used to apply the function to the entire dataset, and the resulting list was compiled back into a data frame for further analysis

### Ranked Character Data

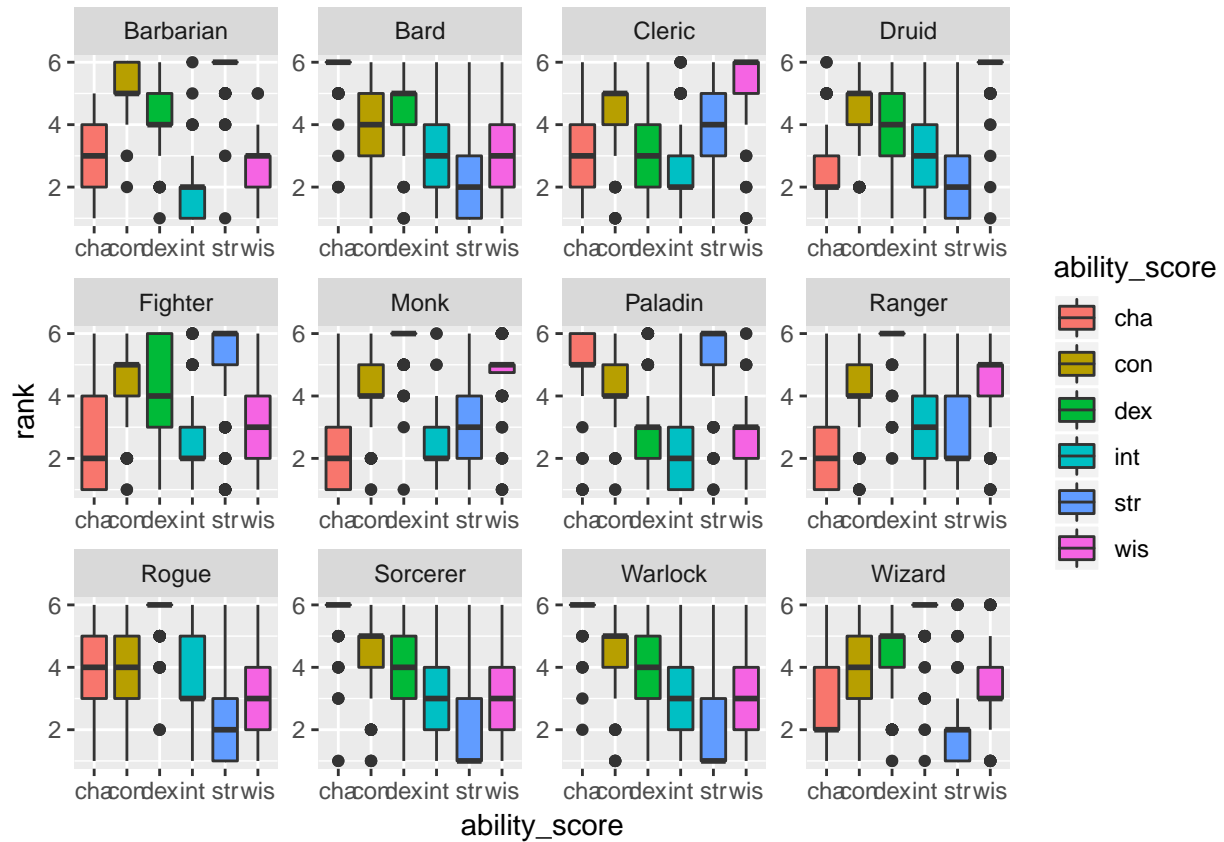
```
##      class      race str dex con int wis cha
## 1  Wizard      Human   1   3   6   6   4   2
## 2  Paladin Dragonborn   6   3   4   3   3   6
## 3   Druid      Human   2   5   5   3   6   2
## 4  Wizard      Human   1   4   4   6   2   5
## 5  Wizard      Human   1   2   3   6   5   5
## 6  Wizard      Human   1   2   3   6   5   5
```

The ranked data was revisualized and the models were re-run to see if there was any improvement.

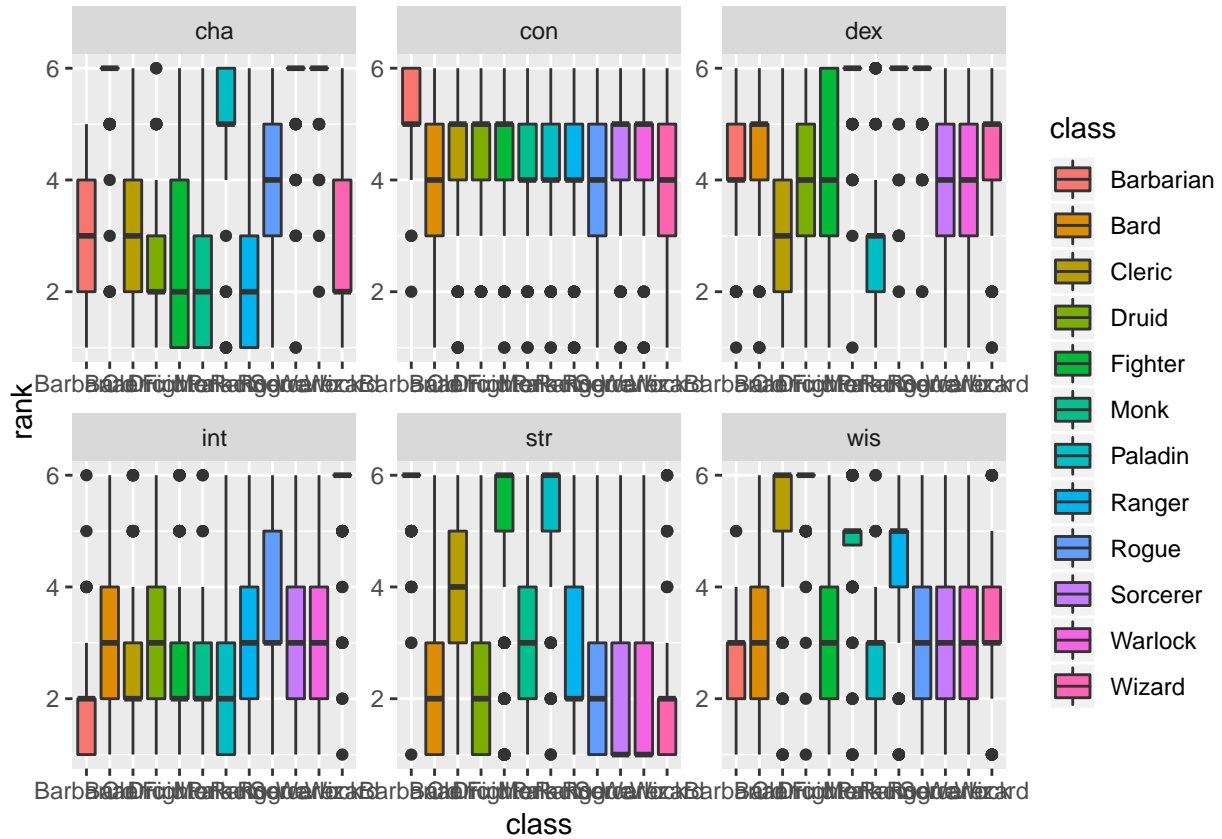
## More Data Visualizations

As expected, the visualizations with the ranked ability scores look almost identical to the visualizations with the raw ability scores, aside from the scales.

### Ranked Ability Scores by Class



### Class by Ranked Ability Scores



## A Third Model

The third model tested was the k-nearest neighbors (kNN) method tuned for k with the ranked ability scores.

### kNN Model

The accuracy of the **ranked tuned kNN** model is 0.49.

## A Fourth Model

The fourth and final model tested was a cp-optimized regression tree with the ranked ability score data.

### Optimized Regression Tree

