

Gotta Predict'em All!

Pokémon Go, a location-based, augmented reality mobile game released by Niantic Inc. The players use the GPS to locate, capture, and battle fictional creatures in a virtual setting who appear on the screen as if they were in the same real-world location as the player.

Like many millennials, I grew up watching Pokémon. So, when this game was released, I was undoubtedly excited, running on the streets of Jaipur looking for the rarest Pokémon's (but always ended up with common Pokémon's)

Although the game was a huge success, there were many flaws that needed to be addressed which ultimately caused the downfall of it.

Dataset:

Dataset consists of roughly 293,000 pokemon sightings (historical appearances of Pokemon), having coordinates, time, weather, population density, distance to pokestops/ gyms etc. as features. The target is to train a machine learning algorithm so that it can predict where pokemon appear in future. So, can you predict'em all?)

Feature description

- pokemonId the identifier of a pokemon, should be deleted to not affect predictions. (numeric; ranges between 1 and 151)
 - https://pokemondb.net/pokedex/national for complete list of all Pokémons ids of gen 1.
- latitude, longitude coordinates of a sighting (numeric)
- appearedLocalTime exact time of a sighting in format yyyy-mm-dd'T'hh-mmss.ms'Z' (nominal)
- cellid 90-5850m geographic position projected on a <u>S2 Cell</u>, with cell sizes ranging from 90 to 5850m (numeric)
- appearedTimeOfDay time of the day of a sighting (night, evening, afternoon, morning)
- appearedHour/appearedMinute local hour/minute of a sighting (numeric)
- appearedDayOfWeek week day of a sighting (Monday, Tuesday, Wednesday, Thursday, Friday, Saturday, Sunday)
- appearedDay/appearedMonth/appearedYear day/month/year of a sighting (numeric)

- terrainType terrain where pokemon appeared described with help of <u>GLCF Modis Land</u> <u>Cover</u> (numeric)
- *closeToWater* did pokemon appear close (100m or less) to water (Boolean, same source as above)
- city the city of a sighting (nominal)
- continent (not always parsed right) the continent of a sighting (nominal)
- weather weather type during a sighting (Foggy Clear, PartlyCloudy, MostlyCloudy,
 Overcast, Rain, BreezyandOvercast, LightRain, Drizzle, BreezyandPartlyCloudy, HeavyRain,
 BreezyandMostlyCloudy, Breezy, Windy, WindyandFoggy, Humid, Dry,
 WindyandPartlyCloudy, DryandMostlyCloudy, DryandPartlyCloudy, DrizzleandBreezy,
 LightRainandBreezy, HumidandPartlyCloudy, HumidandOvercast, RainandWindy)
 // Source for all weather features
- temperature temperature in celsius at the location of a sighting (numeric)
- windSpeed speed of the wind in km/h at the location of a sighting (numeric)
- windBearing wind direction (numeric)
- pressure atmospheric pressure in bar at the location of a sighting (numeric)
- weatherlcon a compact representation of the weather at the location of a sighting (fog, clear-night, partly-cloudy-night, partly-cloudy-day, cloudy, clear-day, rain, wind)
- sunriseMinutesMidnight-sunsetMinutesBefore time of appearance relatively to sunrise/sunset <u>Source</u>
- population density what is the population density per square km of a sighting (numeric, <u>Source</u>)
- *urban-rural* how urban is location where pokemon appeared (Boolean, built on Population density, <200 for rural, >=200 and <400 for midUrban, >=400 and <800 for subUrban, >800 for urban)
- gymDistanceKm, pokestopDistanceKm how far is the nearest gym/pokestop in km from a sighting? (numeric, extracted from this dataset)
- gymIn100m-pokestopIn5000m is there a gym/pokestop in 100/200/etc meters? (Boolean)
- cooc 1-cooc 151 co-occurrence with any other pokemon (pokemon ids range between 1 and 151) within 100m distance and within the last 24 hours (Boolean)
- *class* says which pokemonId it is, to be predicted.

Plot Pokemon activity data points based on latitude and longitude globally on the map with the help of matplotlibs basemap Toolkit.

Sample image output:



- Which cities have the highest Pokémon activities count. Plot a bar graph using seaborn and matplotlib.
- Perform Classification of Pokémon's(id) by continents latitude and longitude.
- Add a column with label 'names' and create a dict of all names from the pokemondb url which corresponds to the ids.

 (https://pokemondb.net/pokedex/national)

 Now using map function, associate the names with the provided ids.
- Predict where a Pokémon of your choice will spawn.

Please make a proper report on this project and submit soft copy to Forsk official email id.

It should atleast contain the following SECTIONS:

- Cover page
- Certificate on Forsk letter head
- Acknowledgement
- Abstract
- List of Figures
- Table of contents
- Introduction
- Background theory/Motivation
- Methodology:

Data Pre-processing / Data Munging

discussion on Regression methods used

discussion on Classification methods used

(knn/logisitic)

Data Visualization.

- Result Analysis
- Conclusion.
- Reference [OPTIONAL]
- ANNEXTURE [WILL CONTAIN THE CODE HERE, don't include lengthy codes in the main part of the report, use code snippets for specific sections]

(Also include the code.py file along with the report in zip folder)