Legendary Pokemon Classification

PRESENTED BY: RAUNAK YADAV

AVINASH CHHIMAL

Topics

In this presentation I am going to talk about following topics of legendary pokemon classification using machine learning, specifically a TensorFlow Artificial neural network (ANN).

The following topics are given below:

Importing Libraries

Loading Data

Data Cleaning and Preprocessing

Exploratory Data Analysis (EDA)

Model Building and Training

Importing Libraries

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import plotly.express as px
From sklearn.preprocessing import LabelEncoder, StandardScaler
From sklearn.model_selection import train test split
import tensorflow as tf
From sklearn.decomposition import PCA
```

Loading Data

data

	#	Name	Type 1	Type 2	Total	НР	Attack	Defense	Sp. Atk	Sp. Def	Speed	Generation	Legendary
0	1	Bulbasaur	Grass	Poison	318	45	49	49	65	65	45	1	False
1	2	lvysaur	Grass	Poison	405	60	62	63	80	80	60	1	False
2	3	Venusaur	Grass	Poison	525	80	82	83	100	100	80	1	False
3	3	VenusaurMega Venusaur	Grass	Poison	625	80	100	123	122	120	80	1	False
4	4	Charmander	Fire	NaN	309	39	52	43	60	50	65	1	False
795	719	Diancie	Rock	Fairy	600	50	100	150	100	150	50	6	True
796	719	DiancieMega Diancie	Rock	Fairy	700	50	160	110	160	110	110	6	True
797	720	HoopaHoopa Confined	Psychic	Ghost	600	80	110	60	150	130	70	6	True
798	720	HoopaHoopa Unbound	Psychic	Dark	680	80	160	60	170	130	80	6	True
799	721	Volcanion	Fire	Water	600	80	110	120	130	90	70	6	True

800 rows × 13 columns

Data Cleaning and Preprocessing

```
data.info()
kclass 'pandas.core.frame.DataFrame'>
RangeIndex: 800 entries, 0 to 799
Data columns (total 13 columns):
                Non-Null Count Dtype
    Column
                 800 non-null
                                 int64
                                object
                800 non-null
    Name
                800 non-null
                                 object
    Type 1
                414 non-null
                                 object
    Type 2
                800 non-null
    Total
                                 int64
                800 non-null
                                 int64
                800 non-null
    Attack
                                 int64
    Defense
                800 non-null
                                 int64
                800 non-null
    Sp. Atk
                                 int64
    Sp. Def
                800 non-null
                                int64
    Speed
                800 non-null
                                 int64
    Generation 800 non-null
                                 int64
                800 non-null
    Legendary
                                 bool
dtypes: bool(1), int64(9), object(3)
memory usage: 75.9+ KB
```

```
In [14]:
           data.dtypes
Dut[14]:
                         object
          Type 1
          Total
                          int64
          HP
                          int64
          Attack
                          int64
          Defense
                          int64
          Sp. Atk
                          int64
          Sp. Def
                          int64
          Speed
                          int64
          Generation
                          int64
          Legendary
                          int64
          dtype: object
```

Exploratory Data Analysis

```
In [10]:
          data['Type 1'].unique()
Out[10]: array(['Grass', 'Fire', 'Water', 'Bug', 'Normal', 'Poison', 'Electric',
                 'Ground', 'Fairy', 'Fighting', 'Psychic', 'Rock', 'Ghost', 'Ice',
                 'Dragon', 'Dark', 'Steel', 'Flying'], dtype=object)
In [11]:
          numeric_columns = data.drop('Type 1', axis=1).columns
In [12]:
          correlation matrix = data[numeric_columns].corr()
          plt.figure(figsize=(18, 15))
          sns.heatmap(correlation matrix, annot=True, vmin=-1.0, vmax=1.0)
          plt.show()
```

Model Building and Training

```
In [21]:
          X.shape
Out[21]: (800, 26)
In [22]:
          inputs = tf.keras.Input(shape=(26,))
          x = tf.keras.layers.Dense(64, activation='relu')(inputs)
          x = tf.keras.layers.Dense(64, activation='relu')(x)
          outputs = tf.keras.layers.Dense(1, activation='sigmoid')(x)
          model = tf.keras.Model(inputs=inputs, outputs=outputs)
          model.compile(
              optimizer='adam',
              loss='binary crossentropy',
              metrics=[
                   'accuracy',
                  tf.keras.metrics.AUC(name='auc')
          batch size = 32
          epochs = 20
          history = model.fit(
              X train,
              y train,
              validation split=0.2,
              batch size=batch size,
              epochs=epochs,
              callbacks=[tf.keras.callbacks.ReduceLROnPlateau()],
              verbose=0
```

Thank you have a good day 💮