

Fungi Edibility Predictive Image Classification Model - Final Notebook

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Imports

```
In [2]: ▶ 1 import csv
          2 import os
          3 os.environ["PROTOCOL_BUFFERS_PYTHON_IMPLEMENTATION"] = "python"
          4 import yaml
          5 import joblib
          6
          7 import pandas as pd
          8 import numpy as np
          9 import pickle
         10 import streamlit as st
         11 import matplotlib.pyplot as plt
         12 import tensorflow as tf
         13 from keras import models
         14 from PIL import Image
         15
         16
         17 from tensorflow.keras import layers
         18 from tensorflow.keras.preprocessing.image import img_to_array, load_image
         19 from tensorflow.keras.utils import to_categorical
         20 from tensorflow.keras.models import load_model
         21 from tensorflow.keras.preprocessing import image
         22 from tensorflow.keras.applications.vgg16 import preprocess_input
         23
         24 from sklearn.pipeline import Pipeline
         25 from sklearn.preprocessing import FunctionTransformer
         26 from sklearn.model_selection import train_test_split
         27 from imblearn.over_sampling import SMOTE
         28 from sklearn.dummy import DummyClassifier
         29 from sklearn.metrics import accuracy_score
         30 from sklearn.preprocessing import LabelEncoder
```

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Loading Images and Labels Files

```
In [ ]: 1 images_flat = np.loadtxt('images.csv', delimiter=',')
        2 images = images_flat.reshape((images.shape[0],) + images.shape[1:])
```

```
In [ ]: 1 labels_encoded = np.loadtxt('labels.csv', delimiter=',')
        2 labels = label_encoder.inverse_transform(labels_encoded)
```

```
In [ ]: 1 print(images.shape)
        2 print(labels_decoded.shape)
```

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Visualizing How the Model Classifies Images

Seeing what the most important features of an image is when it comes to clasifying. Going to be trying various viusal techniques.

Visualizing Activations

This is a function written to view what different activation layers of my model are, this will help understand which parts of the mushroom are important for classification.

```
In [29]: 1 def visualize_activations(model, images):
        2
        3     layer_outputs = [layer.output for layer in model.layers]
        4     activation_model = models.Model(inputs=model.input, outputs=layer_outputs)
        5
        6     for image in images:
        7
        8         activations = activation_model.predict(np.expand_dims(image, axis=0))
        9
        10
        11         for layer_activation in activations:
        12
        13             for i in range(layer_activation.shape[-1]):
        14                 plt.imshow(layer_activation[0, :, :, i], cmap='jet')
        15                 plt.show()
```

Visualizing Predictions

This function was written to understand how the model is predicting an images class, and the probability of it being correct, this may be useful when determining how well the model is working.

```
In [30]: 1 def visualize_predictions(model, images):
2         for image in images:
3
4             predictions = model.predict(np.expand_dims(image, axis=0))
5             predicted_class = np.argmax(predictions)
6             class_probability = np.max(predictions)
7
8
9             plt.imshow(image)
10            plt.title(f'Predicted class: {predicted_class}, Probability:
11            plt.axis('off')
12            plt.show()
```

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Creating variables with smaller subsets of images to view a wider variety data.

```
In [33]: 1 images_p1 = images[:400]
```

```
In [34]: 1 images_p2 = images[400:900]
```

```
In [35]: 1 images_p3 = images[900:1200]
```

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Redefining my Model's Shape

This is so my visualization functions will work.

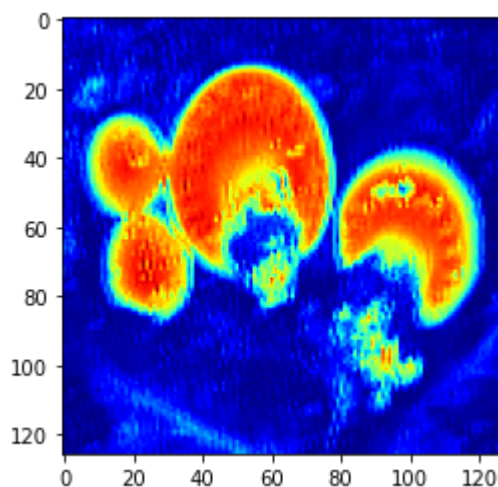
```
In [36]: 1 model = tf.keras.Sequential([
2         layers.Conv2D(32, (3, 3), activation='relu', input_shape=(28, 28, 3)),
3         layers.SeparableConv2D(filters=64, kernel_size=(3, 3), activation='relu'),
4         layers.MaxPooling2D((2, 2)),
5         layers.Conv2D(filters=128, kernel_size=(3, 3), activation='relu'),
6         layers.GlobalAveragePooling2D(),
7         layers.Flatten(),
8         layers.Dense(64, activation='relu'),
9         layers.Dense(10, activation='softmax')
10      ])
11
12 early_stopping = tf.keras.callbacks.EarlyStopping(
13     monitor="val_loss",
14     patience=4,
15     verbose=1,
16     restore_best_weights=True
17 )
```

Visualizing Activations

First out of three image subsets

```
In [37]: 1 visualize_activations(model, images_p1)
```

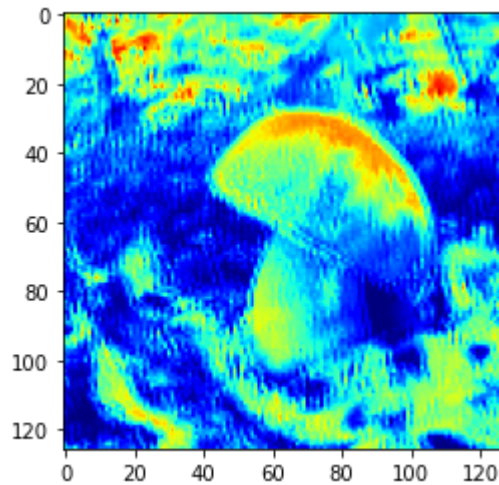
WARNING:tensorflow:Model was constructed with shape (None, 28, 28, 3) for input Tensor("conv2d_3_input:0", shape=(None, 28, 28, 3), dtype=float32), but it was called on an input with incompatible shape (None, 128, 128, 3).



Second image subset

In [41]: 1 visualize_activations(model, images_p2)

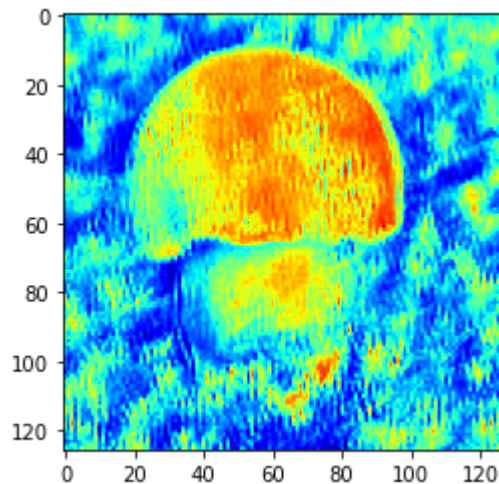
WARNING:tensorflow:Model was constructed with shape (None, 28, 28, 3) for input Tensor("conv2d_3_input:0", shape=(None, 28, 28, 3), dtype=float32), but it was called on an input with incompatible shape (None, 128, 128, 3).



Third image subset

In [42]: 1 visualize_activations(model, images_p3)

WARNING:tensorflow:Model was constructed with shape (None, 28, 28, 3) for input Tensor("conv2d_3_input:0", shape=(None, 28, 28, 3), dtype=float32), but it was called on an input with incompatible shape (None, 128, 128, 3).



Visualizing Predictions

First out of three image subsets

In [45]: 1 visualize_predictions(model, images_p1)

Predicted class: 6, Probability: 0.10



Predicted class: 6, Probability: 0.10



Second image subset

In [46]: 1 visualize_predictions(model, images_p2)

Predicted class: 6, Probability: 0.10



Predicted class: 6, Probability: 0.10



Third image subset

```
In [47]: 1 visualize_predictions(model, images_p3)
```

Predicted class: 6, Probability: 0.10



Predicted class: 6, Probability: 0.10



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Deploying the Model

```
In [ ]: 1
```

```
In [ ]: 1
```

```
In [ ]: 1 #pickle_out = open("classifier.pkl", "wb")  
2 #pickle.dump(model, pickle_out)  
3 #pickle_out.close()
```

```
In [ ]: 1
```

```

In [ ]: ▶ 1 #loaded_model = load_model("classifier.h5")
2 model_path = "classifier.pkl"
3
4 @st.cache(hash_funcs={tf.keras.models.Model: lambda _: None})
5 def load_model():
6     return tf.keras.models.load_model(model_path)
7
8 loaded_model = load_model()
9
10
11 def welcome():
12     return 'welcome fellow msuhroom enthusiasts'
13
14 def predictions(image):
15     model = loaded_model
16     prediction = model.predict(image)
17     predicted_class = np.argmax(prediction)
18     predicted_label = labels[predicted_class]
19     return predictions
20
21
22 def main():
23     st.title("Mushroom Edibility Prediction")
24
25     html_temp = """
26     <div style = "background-color:yellow;padding:13px">
27     <h1 style = "color:black;text-align:center;">Streamlit Iris Flower
28     </div>
29     """
30
31     st.markdown(html_temp, unsafe_allow_html = True)
32
33     uploaded_file = st.file_uploader("Upload a Mushroom Image", type=
34
35     if uploaded_file is not None:
36
37         image = Image.open(uploaded_file)
38         st.image(image, caption='Uploaded Mushroom Image', use_colum
39
40
41         if st.button('Predict'):
42             predicted_label = predict_image(image)
43             st.success(f"Predicted Mushroom Edibility: {predicted_lat
44
45 if __name__ == '__main__':
46     main()

```

```

In [ ]: ▶ 1

```