Week 4 Homework 2: Classification on Colab using MNIST dataset

GitHub link: https://github.com/kathyshe/Machine-Learning/tree/main/Supervised-Learning/Classification%20on%20Colab%20using%20MNIST%20dataset

Introduction

The MNIST database is a large database of handwritten digits that is commonly used for training various image processing systems. This dataset is one of the most common datasets used for image classification and accessible from many different sources.

MNIST dataset, which is a set of 70,000 small images of digits handwritten by high school students and employees of the US Census Bureau. Each image is labeled with the digit it represents. This set is called the "hello world" of Machine Learning: Whenever people come up with a new classification algorithm they are curious to see how it will perform on MNIST.

Scikit-Learn provides many helper functions to download popular datasets. MNIST is one of them.

Google Colab provides free access to a Jupyter Notebook and a small version of the MNIST dataset. Aspiring data scientists can use the platform to practice and experiment with the MNIST dataset. The advantage of using Google Colab is that many popular Python libraries are already pre-installed, making it easy to get started with deep learning and computer vision. The MNIST dataset is also readily available as a sample, allowing users to quickly dive into data science and machine learning.

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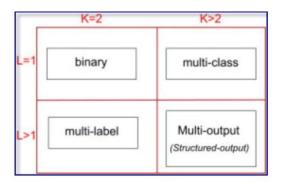
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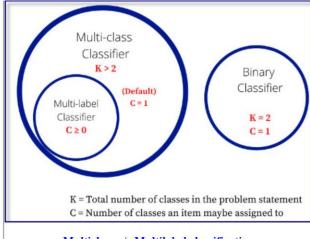
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Scalable multi-output label prediction: From classifier chains to classifier trellises



Multiclass v/s Multilabel classification

Implementation on Google Colab:

https://colab.research.google.com/drive/1nhHVH-gTQnfi3rv7ZS-mjmm8y8Za7uqo?usp=sharing

Execute classification.ipynb:

https://github.com/divyapandey03/Machine-Learning/blob/main/Supervised% 20 Learning/Classification% 20 on% 20 Colab% 20 using% 20 MNIST% 20 dataset/classification.ipynb

Output:

```
[6] %matplotlib inline
import matplotlib as mpl
import matplotlib.pyplot as plt

some_digit = X[0]
some_digit_image = some_digit.reshape(28, 28)
plt.imshow(some_digit_image, cmap=mpl.cm.binary)
plt.axis("off")

save_fig("some_digit_plot")
plt.show()

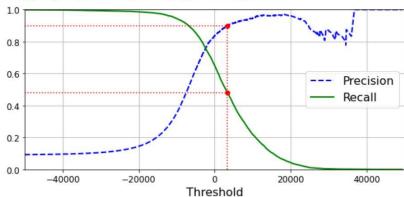
Saving figure some_digit_plot

[7] y[0]

'5'
```

```
def plot_precision_recall_vs_threshold(precisions, recalls, thresholds):
         plt.plot(thresholds, precisions[:-1], "b--", label="Precision", linewidth=2)
         plt.plot(thresholds, recalls[:-1], "g-", label="Recall", linewidth=2)
         plt.legend(loc="center right", fontsize=16) # Not shown in the book
                                                  # Not shown
         plt.xlabel("Threshold", fontsize=16)
         plt.grid(True)
                                                    # Not shown
         plt.axis([-50000, 50000, 0, 1])
                                                    # Not shown
     recall_90_precision = recalls[np.argmax(precisions >= 0.90)]
     threshold\_90\_precision = thresholds[np.argmax(precisions >= 0.90)]
     plt.figure(figsize=(8, 4))
                                                                                                 # Not shown
     plot_precision_recall_vs_threshold(precisions, recalls, thresholds)
     plt.plot([threshold_90_precision, threshold_90_precision], [0., 0.9], "r:")
                                                                                                # Not shown
     plt.plot([-50000, threshold_90_precision], [0.9, 0.9], "r:")
     plt.plot([-50000, threshold_90_precision], [recall_90_precision, recall_90_precision], "r:")# Not shown
     plt.plot([threshold_90_precision], [0.9], "ro")
     plt.plot([threshold_90_precision], [recall_90_precision], "ro")
                                                                                                 # Not shown
                                                                                                # Not shown
     save_fig("precision_recall_vs_threshold_plot")
     plt.show()
```

Saving figure precision_recall_vs_threshold_plot



```
def plot_precision_vs_recall(precisions, recalls):
    plt.plot(recalls, precisions, "b-", linewidth=2)
    plt.xlabel("Recall", fontsize=16)
    plt.ylabel("Precision", fontsize=16)
    plt.axis([0, 1, 0, 1])
    plt.grid(True)

plt.figure(figsize=(8, 6))
    plot_precision_vs_recall(precisions, recalls)
    plt.plot([recall_90_precision, recall_90_precision], [0., 0.9], "r:")
    plt.plot([0.0, recall_90_precision], [0.9, 0.9], "r:")
    plt.plot([recall_90_precision], [0.9], "ro")
    save_fig("precision_vs_recall_plot")
    plt.show()
```

Saving figure precision_vs_recall_plot

