```
##reading one image and viewing
image = imread("/kaggle/input/mnist-digit/MNIST Dataset JPG format/MNIST - JPG - training/0/1.jpg")
imshow(image)
print(f"The shape of image is {image.shape}")

The shape of image is (28, 28)
<ipython-input-29-dc33dd558c6e>:3: FutureWarning: `imshow` is deprecated since version 0.25 and will be removed in version 0.27. Please use `matpl
imshow(image)

0-

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```

# You can write up to 20GB to the current directory (/kaggle/working/) that gets preserved as output when you create a version using "Save & Run All"

# This Python 3 environment comes with many helpful analytics libraries installed
# It is defined by the kaggle/python Docker image: https://github.com/kaggle/docker-python

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image\_path = os.path.join(label\_path, image\_file)

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##Here, I am defining the function to load the images from folder and simulteniously computing hog features

for image\_file in tqdm(os.listdir(label\_path), desc=f"Processing class {label}"):

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label\_path = os.path.join(dataset\_path, label)

image = imread(image\_path)

feature = compute\_hog\_features(image)

## to create dataset for my ml model.
def load\_dataset(dataset\_path):

for label in os.listdir(dataset\_path):

if os.path.isdir(label\_path):

# For example, running this (by clicking run or pressing Shift+Enter) will list all files under the input directory

# You can also write temporary files to <a href="https://kaggle/temp/">kaggle/temp/</a>, but they won't be saved outside of the current session

import pandas as pd # data processing, CSV file I/O (e.g. pd.read\_csv)
# Input data files are available in the read-only "../input/" directory

from skimage.io import imread ##importing the required libraries

# For example, here's several helpful packages to load

import numpy as np # linear algebra

from skimage.color import rgb2gray
from skimage.transform import resize
from scipy.ndimage import sobel
from sklearn.svm import SVC

from tqdm import tqdm

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features = []
labels = []

from sklearn.metrics import accuracy\_score

from sklearn.model\_selection import train\_test\_split

import os

```
##loading the train dataset from training folder
train_path="/kaggle/input/mnist-digit/MNIST Dataset JPG format/MNIST - JPG - training"
X_train, y_train = load_dataset(train_path)
→ Processing class 7: 100%
                                           6265/6265 [00:40<00:00, 153.30it/s]
     Processing class 2: 100%
                                           5958/5958 [00:39<00:00, 152.02it/s]
     Processing class 5: 100%
                                           5421/5421 [00:35<00:00, 152.81it/s]
     Processing class 8: 100%
                                           5851/5851 [00:36<00:00, 159.72it/s]
     Processing class 0: 100%
                                           5923/5923 [00:37<00:00, 157.35it/s]
     Processing class 3: 100%
                                           6131/6131 [00:38<00:00, 157.95it/s]
     Processing class 1: 100%
                                           6742/6742 [00:42<00:00, 157.96it/s]
     Processing class 4: 100%
                                           5842/5842 [00:37<00:00, 157.54it/s]
     Processing class 9: 100%
                                           5949/5949 [00:38<00:00, 156.01it/s]
     Processing class 6: 100%
                                           5918/5918 [00:38<00:00, 155.22it/s]
##loading the dataset from test folder
test_path="/kaggle/input/mnist-digit/MNIST Dataset JPG format/MNIST - JPG - testing"
X_test,y_test = load_dataset(test_path)
→ Processing class 7: 100%
                                           1028/1028 [00:06<00:00, 158.43it/s]
     Processing class 2: 100%
                                           1032/1032 [00:06<00:00, 150.37it/s]
                                           892/892 [00:05<00:00, 157.53it/s]
     Processing class 5: 100%
     Processing class 8: 100%
                                           974/974 [00:06<00:00, 149.23it/s]
     Processing class 0: 100%
                                           980/980 [00:06<00:00, 153.40it/s]
                                           1010/1010 [00:06<00:00, 159.59it/s]
     Processing class 3: 100%
     Processing class 1: 100%
                                           1135/1135 [00:07<00:00, 156.09it/s]
     Processing class 4: 100%
                                           982/982 [00:06<00:00, 155.85it/s]
     Processing class 9: 100%
                                           1009/1009 [00:06<00:00, 153.70it/s]
     Processing class 6: 100%
                                           958/958 [00:05<00:00, 160.17it/s]
## function to extract the hog(Histogram of Oriented Gradients) features
def compute_hog_features(image, pixels_per_cell=(8, 8), cells_per_block=(2, 2), orientations=9):
    # Convert to grayscale and resize for uniformity
    image = resize(image, (64, 64)) # Resize all images to 64x64 for consistency
    # Computing the gradients using sobel filter
    gx = sobel(image, axis=1)
    gy = sobel(image, axis=0)
    magnitude = np.sqrt(gx**2 + gy**2)
                                           ## computing the magnitude of gradient
    orientation = (np.arctan2(gy, gx) * (180 / np.pi)) % 180 ## computing the angle(orientation) of gradient
    # Calculating the HOG features
    cell_rows, cell_cols = pixels_per_cell
    block_rows, block_cols = cells_per_block
    height, width = image.shape
    num_cells_x = width // cell_cols
    num_cells_y = height // cell_rows
    histograms = np.zeros((num_cells_y, num_cells_x, orientations))
    bin_edges = np.linspace(0, 180, orientations + 1)
    for i in range(num_cells_y):
        for j in range(num_cells_x):
            \texttt{cell\_magnitude} = \texttt{magnitude}[\texttt{i} * \texttt{cell\_rows:}(\texttt{i} + \texttt{1}) * \texttt{cell\_rows}, \texttt{j} * \texttt{cell\_cols:}(\texttt{j} + \texttt{1}) * \texttt{cell\_cols}]
            cell_orientation = orientation[i * cell_rows:(i + 1) * cell_rows, j * cell_cols:(j + 1) * cell_cols]
                        _ = np.histogram(cell_orientation, bins=bin_edges, weights=cell_magnitude)
            histogram,
            histograms[i, j, :] = histogram
    block_features = []
    for i in range(num_cells_y - block_rows + 1):
        for j in range(num_cells_x - block_cols + 1):
            block = histograms[i:i + block_rows, j:j + block_cols, :].ravel()
            block = block / np.sqrt(np.sum(block**2) + 1e-6)
            block_features.append(block)
    return np.concatenate(block features)
##Loading the SVC model for classification
from sklearn.svm import SVC
from sklearn.metrics import accuracy_score
svm = SVC(kernel='rbf', C=1.0)
svm.fit(X_train, y_train)
Show hidden output
# Evaluating the model
y_pred = svm.predict(X_test)
```

accuracy = accuracy\_score(y\_test, y\_pred)

→ The accuracy of model is: 99.11%

print(f"The accuracy of model is: {accuracy \* 100:.2f}%")

from sklearn.metrics import classification\_report
report=classification\_report(y\_test,y\_pred)

## print(report)

₹	precision	recall	f1-score	support
0	0.99	1.00	1.00	980
1	0.99	1.00	0.99	1135
2	0.99	0.99	0.99	1032
3	0.99	0.99	0.99	1010
4	0.99	0.99	0.99	982
5	0.99	0.99	0.99	892
6	0.99	0.99	0.99	958
7	0.99	0.99	0.99	1028
8	0.99	0.99	0.99	974
9	0.99	0.98	0.98	1009
accuracy			0.99	10000
macro avg	0.99	0.99	0.99	10000
weighted avg	0.99	0.99	0.99	10000

Start coding or generate with AI.