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The Project Report for the project entitled "Devanāgarī Numerals Recognition System"

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Devanāgarī Numerals Recognition System

1. Abstract

This project, "Devanāgarī Numerals Recognition System", developed by Umang Goswami, aims to create a robust and reliable system for the recognition of Devanāgarī numerals. The is to develop a highly accurate and efficient Devanāgarī numeral recognition system that can be seamlessly integrated into a wide range of applications, ultimately enhancing the processing and management of Devanāgarī-based data.

The system leverages state-of-the-art machine learning algorithms to accurately identify hand-written Devanāgarī numerals, enabling highly efficient and precise data processing in various applications that necessitate Devanāgarī numeral recognition. These applications span diverse domains, including document processing, financial data analysis, and information extraction from Devanāgarī-based documents.

By harnessing advanced computational techniques, this innovative system promises to significantly enhance the accuracy and productivity of tasks involving Devanāgarī numeral recognition, contributing to improved data management and analysis across various industries and research fields.

2. Technologies Used

The Devanāgarī Numerals Recognition System utilizes a combination of advanced machine learning techniques and image processing algorithms to achieve high-accuracy recognition of Devanāgarī numerals. Specifically, the system employs Convolutional Neural Networks as the primary deep learning architecture for image classification.

The neural networks are trained on a large and diverse dataset of Devanāgarī numeral images, comprising a total of 26,500 samples. This extensive

training enables the system to learn the distinctive features and patterns associated with each numeral, allowing for highly accurate recognition of hand-written Devanāgarī numerals. The system's performance has been extensively evaluated, demonstrating its robustness and reliability in a variety of real-world applications.

In brief various technologies used for the development of the aforementioned project are as follows:

- **IDE:** Pycahrm 2024.1 (Professional Edition)
- **Programming Language:** Python v. 3.12
- Various Libraries used: scikit-learn, OpenCV, TensorFlow, NumPy, SciPy, Keras, pillow, etc. For the complete list of libraries use see requirements.txt at next page.

Specifications of the computer systems used for development of the project are:

- Processor: 12th Gen Intel(R) Core(TM) i7-12650H @ 2.30 GHz
- Installed RAM: 16.0 GB (15.7 GB usable)
- Internal Storage: 1 GB SSD
- System type: 64-bit operating system, x64-based processor
- **OS:** Windows 11 Home Single Language, version 23H2

3. All Libraries used – requirements.txt

```
abs1-py==2.1.0
astunparse==1.6.3
blinker==1.9.0
certifi==2024.8.30
charset-normalizer==3.4.0
click==8.1.7
colorama==0.4.6
Flask==3.1.0
Flask-Cors==5.0.0
flatbuffers==24.3.25
gast==0.6.0
google-pasta==0.2.0
grpcio==1.68.1
h5py==3.12.1
idna==3.10
itsdangerous==2.2.0
Jinja2==3.1.4
joblib==1.4.2
keras==3.7.0
libclang==18.1.1
Markdown==3.7
markdown-it-py==3.0.0
MarkupSafe==3.0.2
mdurl==0.1.2
ml-dtypes==0.4.1
namex==0.0.8
numpy==2.0.2
opency-python==4.10.0.84
opt einsum==3.4.0
optree==0.13.1
packaging==24.2
protobuf==5.29.1
Pygments==2.18.0
requests==2.32.3
rich==13.9.4
scikit-learn==1.5.2
scipy==1.14.1
setuptools==75.6.0
six = 1.17.0
```

```
tensorboard==2.18.0

tensorboard-data-server==0.7.2

tensorflow==2.18.0

tensorflow_intel==2.18.0

termcolor==2.5.0

threadpoolctl==3.5.0

typing_extensions==4.12.2

urllib3==2.2.3

Werkzeug==3.1.3

wheel==0.45.1

wrapt==1.1
```

4. System Requirements

The system requirements for the successful installation and operation of the Devanāgarī Numerals Recognition System are as follows:

4.1. Hardware Requirements:

- Processor: The system requires at least an Intel Core i5 processor or an
 equivalent, with a clock speed of 2.0 GHz or higher. This level of
 processing power is necessary to handle the computationally intensive
 tasks involved in the recognition and classification of the Devanāgarī
 numerals.
- RAM: A minimum of 8 GB of RAM is recommended for the smooth operation of the system. This amount of memory ensures that the system can efficiently load and process the large datasets required for training and running the machine learning models.
- Storage: The system requires 25 GB of available disk space to accommodate the installation files, the training dataset, and any additional resources needed for the recognition process.
- Display: A display with a resolution of 1080p or higher is recommended for optimal visualization and interaction with the system's interface.

4.2. Software Requirements:

- Operating System: The Devanāgarī Numerals Recognition System is designed to operate on at least Windows 10 or Windows 11 operating systems, or their equivalent. These modern operating systems provide the necessary software environment and libraries to support the system's functionality.
- Python: The system requires Python version 3.12 or higher to be installed, as the codebase and supporting libraries are developed using this programming language. This version of Python ensures compatibility with the latest advancements in machine learning and image processing libraries.

5. Dataset used for training

The Devanāgarī Numerals Recognition System was developed using a comprehensive dataset of 26,500 Devanāgarī numeral images. I had modified/restructured the dataset, its file structures as to better suit my project needs and feasibility. The dataset was obtained from the following source:

```
Handwritten Devanāgarī characters - vowels and numerals
(38,750 isolated images + 9,200 isolated images). (2023).
[Dataset]. In Mendeley Data (Version 4).
doi: 10.17632/pxrnvp4yy8.4 Retrived from
https://data.mendeley.com/datasets/pxrnvp4yy8/4
```

whose description may be found in the research paper –

Prashanth, D. S., Mehta, R. V. K., & Challa, N. P. (2021). A multi-purpose dataset of Devanāgarī script comprising of isolated numerals and vowels. Data in Brief, 40, 107723. https://doi.org/10.1016/j.dib.2021.107723 Retrived from https://pdf.sciencedirectassets.com/311593/1-s2.0-s2352340921009987/main.pdf

6. Source Code

6.1. main.py

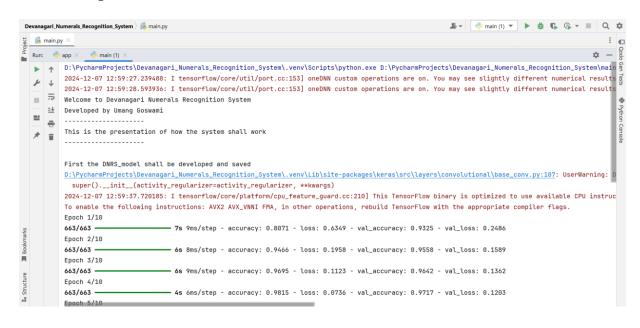
```
import os
from src.data loader import Load Data, Preprocess Data
from src.model import Create Model
from src.train import Train Model
from src.evaluate import Evaluate Model
from src.predict import Predict Character
from sklearn.model selection import train test split
from tkinter import *
from tkinter.filedialog import askopenfilename
root = Tk()
data dir = os.path.abspath(r".\dataset") # Path to the
dataset folder
def prepare DNRS model():
    # Load and preprocess the data
    X, y = Load Data(data dir)
   X, y = Preprocess_Data(X, y)
    # Split the dataset into training and testing sets
   X train, X test, y train, y test = train test split(X, y,
test_size=0.2, random state=42)
    # Create the model
   model = Create Model(y.shape[1])
    # Train the model
    Train Model(model, X train, y train, X test, y test)
    # Evaluate the model
    Evaluate Model(model, X test, y test)
    return model
def make label predictions(model):
    Tk.lift(root)
    img path = str(askopenfilename(defaultextension=".img",
initialdir=".",
          title="Select an image to make label-predictions"))
    return Predict Character(model=model, image path=img path)
def main():
   print("Welcome to Devanagari Numerals Recognition System")
   print("Developed by Umang Goswami")
```

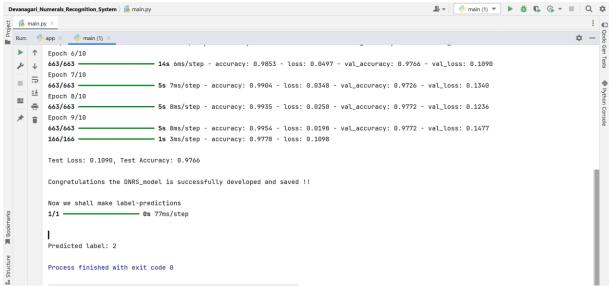
```
print("-" * 21)
   print("This is the presentation of how the system shall
work")
   print("-" * 21)
   print("\nFirst the DNRS model shall be developed and
saved")
    DNRS model = prepare DNRS model()
    print("\nCongratulations the DNRS model is successfully
developed and saved !!")
   print("\nNow we shall make label-predictions")
   predicted label = make label predictions(DNRS model)
   print(f'\nPredicted label: {predicted label}')
if _name__ == '__main__':
   main()
6.2. src/data loader.py
import os
import cv2
import numpy as np
def Load Data(data dir):
    """Load images and labels from the dataset directory."""
    images = []
    labels = []
    # Iterate through each subfolder in the dataset directory
    for label in os.listdir(data dir):
       label dir = os.path.join(data dir, label)
       if os.path.isdir(label dir):
          for image file in os.listdir(label dir):
             image path = os.path.join(label dir, image file)
             image = cv2.imread(image path,
cv2.IMREAD GRAYSCALE)
             image = cv2.resize(image, (28, 28)) # Resize to
28x28
             if image is not None:
                images.append(image)
                labels.append(int(label)) # Convert label to
integer
    return np.array(images), np.array(labels)
def Preprocess_Data(X, y):
    """Preprocess the images and labels."""
    # Resize images to 28x28 and normalize
    X resized = np.array([cv2.resize(img, (28, 28)) for img in
```

```
X1)
    X normalized = X resized.astype('float32') / 255.0
    X normalized = X normalized.reshape(-1, 28, 28, 1)
Reshape for CNN
    # One-hot encode the labels
    from tensorflow.keras.utils import to categorical
    y encoded = to categorical(y)
    return X normalized, y encoded
6.3. src/ evaluate.py
from data loader import Load Data, Preprocess Data
import os
data dir = os.path.abspath(r".\dataset") # Update this path
X, y = Load Data(data dir)
X, y = Preprocess Data(X, y)
def Evaluate Model(model, X_test, y_test):
    """Evaluate the model on the test data."""
    loss, accuracy = model.evaluate(X test, y test)
   print(f'\nTest Loss: {loss:.4f}, Test Accuracy:
{accuracy:.4f}')
6.4. src/model.py
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Conv2D, MaxPooling2D,
Flatten, Dense
def Create Model(num_classes):
   model = Sequential()
    model.add(Conv2D(32, (3, 3), activation='relu',
input shape=(28, 28, 1)))
   model.add(MaxPooling2D(pool size=(2, 2)))
   model.add(Flatten())
   model.add(Dense(128, activation='relu'))
   model.add(Dense(num classes, activation='softmax'))
    # Compile the model
   model.compile(optimizer='adam',
loss='categorical crossentropy', metrics=['accuracy'])
    return model
6.5. src/predict.py
```

```
import cv2
import numpy as np
def Load and Preprocess Image(image path):
    """Load and preprocess the image for prediction."""
    image = cv2.imread(image path, cv2.IMREAD GRAYSCALE)
    image = cv2.resize(image, (28, 28))
    image = image.astype('float32') / 255.0
    image = image.reshape(-1, 28, 28, 1) # Reshape for the
model
    return image
def Predict Character(model, image path):
    """Predict the character in the given image."""
    image = Load and Preprocess Image(image path)
   prediction = model.predict(image)
   predicted label = np.argmax(prediction)
    return predicted label
6.6. src/train.py
from tensorflow.keras.callbacks import EarlyStopping
def Train Model (model, X train, y train, X test, y test,
epochs=10, batch size=32):
    """Train the model on the training data."""
    # Early stopping to prevent overfitting
    early stopping = EarlyStopping(monitor='val loss',
patience=3, restore best weights=True)
    # Train the model
   model.fit(X train, y train, validation data=(X test,
y_test),
          epochs=epochs, batch size=batch size,
callbacks=[early stopping])
    # Save the model after training
   model.save('Devanagari Numerals Recognition Model.keras')
```

7. Output





8. References

- Handwritten Devanāgarī characters vowels and numerals (38,750 isolated images + 9,200 isolated images). (2023). [Dataset]. In *Mendeley Data* (Version 4). doi:10.17632/pxrnvp4yy8.4 Retrived from https://data.mendeley.com/datasets/pxrnvp4yy8/4
- Prashanth, D. S., Mehta, R. V. K., & Challa, N. P. (2021). A multipurpose dataset of Devanāgarī script comprising of isolated numerals and vowels. *Data in Brief*, 40, 107723. https://doi.org/10.1016/j.dib.2021.107723 Retrived from https://pdf.sciencedirectassets.com/311593/1-s2.0-s2352340921009987/main.pdf