

EFFICIENT SATELLITE IMAGE REPRESENTATION USING DEEP AUTOENCODERS

By Group 28 :

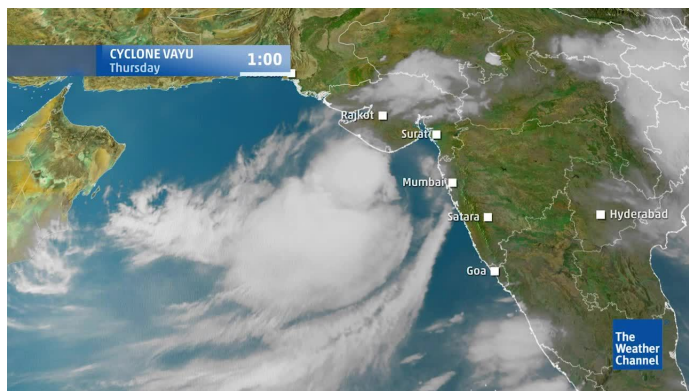
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Importance of Satellite Images

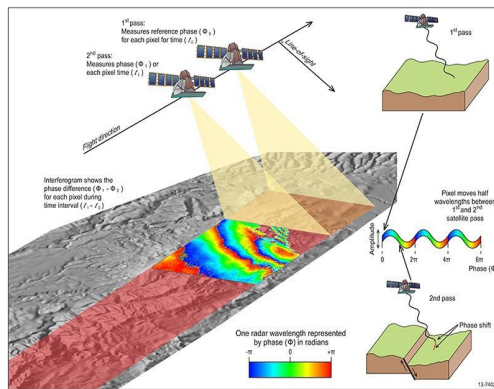
Satellite images give a smooth **representation** of happenings around the **globe** and the **opening** of different **events** in the sky. **Satellite** data consists of **high-resolution** images, hence the data is **high** in **volume**, and needs a **high** amount of **storage**.

How **IMPORTANT** are satellite images for the world ?

- **Weather Forecasting :**
Monitor weather conditions and **parameters** such as low clouds, thunderstorms and even track the evolution of dust plumes in the atmosphere.
- **Geographic Information System (GIS) :**
GIS gathers lots of satellite image data to **analyze** and **extract** useful information from this data like **population** characteristics, **economic development**, etc.
- **InSAR (Interferometric Synthetic Aperture radar) :**
It **predicts** and measures the length of the **displacement** of potential landslides, irrigation, and soil moisture tracking and management.
- **Urban Planning, Crop Prediction, Military Surveillance, Disaster Management and Natural Resource Utilization.**



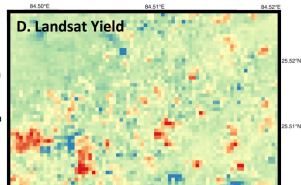
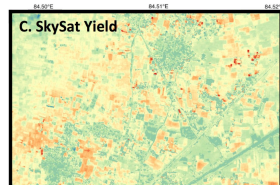
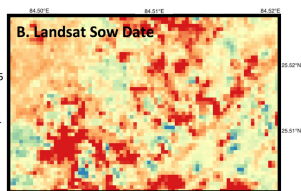
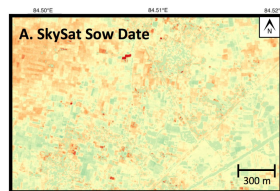
Weather Forecasting



InSAR



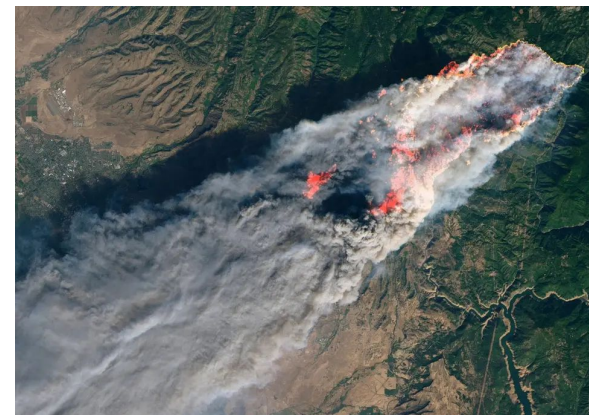
Urban Planning



Crop Prediction



Military Surveillance



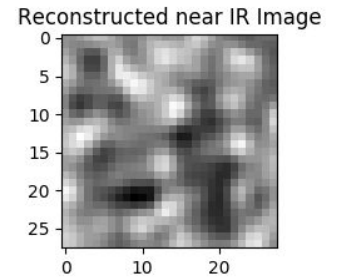
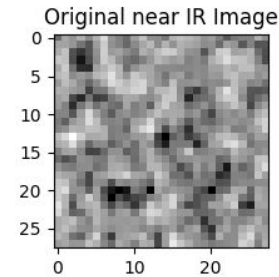
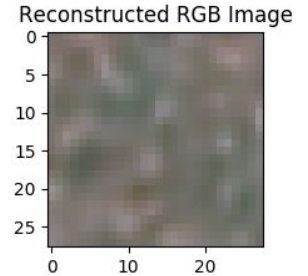
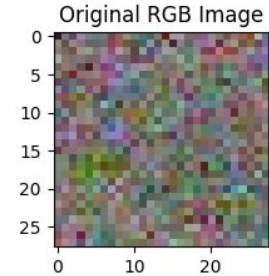
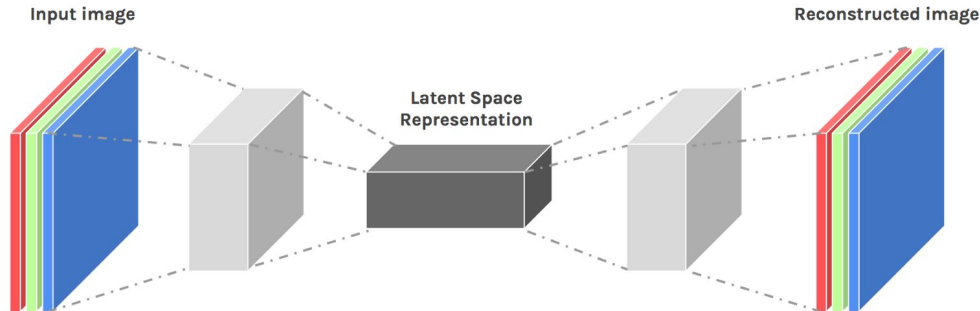
Disaster Management

Autoencoder Theory

Autoencoder : ANN in which the input and output are the same.

Uses :

1. Finding compressed representation of data
2. Dimensionality Reduction
3. Denoising data
4. Data Generation



Model Architecture

Hyperparameters :

Number of Epochs: 30

Loss Used: Mean Square Error

Metrics Used: PSNR, R²

Batch Size: 32

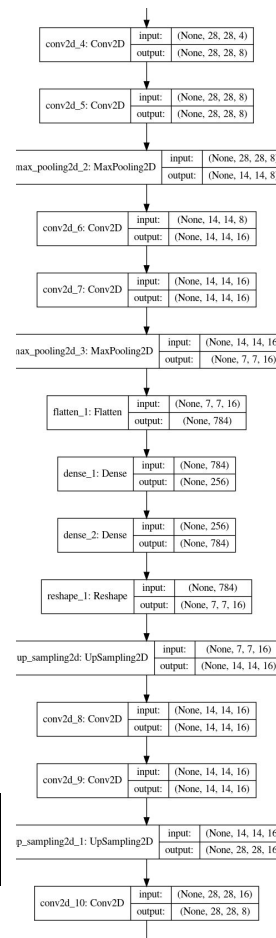
Optimizer Used: Adam

Table - I : Variation of Test Loss, Accuracy and PSNR with size of Latent Space

| S.No. | Size of latent space | Test Loss | Test accuracy | PSNR |
|-------|----------------------|-----------|---------------|--------|
| 1. | 64 | 0.0035 | 91.62% | 30.112 |
| 2. | 128 | 0.0028 | 92.73% | 31.198 |
| 3. | 256 | 0.0023 | 93.57% | 31.933 |
| 4. | 512 | 0.0023 | 92.93% | 31.70 |
| 5. | 1024 | 0.0022 | 93.15 | 31. |

Table - II : Performance of Models for Image Classification

| S.No. | Model Used | Accuracy |
|-------|---------------------------------------|----------|
| 1. | Autoencoder+Logistic Regression | 91% |
| 2. | Autoencoder+Support Vector Classifier | 97% |
| 3. | Autoencoder+Dense Neural Network | 99% |
| 4. | Autoencoder+K-Nearest Neighbor | 70% |
| 5. | Dense Neural Network | 72% |
| 6. | Deep Convolutional Neural Network | 99% |



Source Code : <https://github.com/vaibagga/Satellite-Image-Similarity>

Demonstration Using Grayscale Images

Source Code : <https://github.com/tanishq9/Unsupervised-Image-Retrieval-API>

For demonstration purposes we have used 28*28 single channel grayscale images to easily convey the idea of finding K-most similar images to a given image by building a Image Retrieval System by using Convolutional Autoencoder and Nearest Neighbor technique. The dataset that which we used was **MNIST dataset**, it had **70,000** images out of which **60,000** images were used for training and **10,000** images were used for validation purposes. The model has been trained for 2 epochs having a batch size of 32. Adam optimizer was used for optimization.

autoencoder.py

```
1 input_img = Input(shape=(28,28,1))
2 x = Conv2D(16,(3,3), activation='relu', padding='same')(input_img)
3 x = MaxPooling2D((2,2), padding='same')(x)
4 x = Conv2D(8,(3,3), activation='relu', padding='same')(x)
5 x = MaxPooling2D((2,2), padding='same')(x)
6 x = Conv2D(8,(3,3), activation='relu', padding='same')(x)
7 encoded = MaxPooling2D((2,2), padding='same', name='encoder')(x)
8
9 x = Conv2D(8, (3, 3), activation='relu', padding='same')(encoded)
10 x = UpSampling2D((2, 2))(x)
11 x = Conv2D(8, (3, 3), activation='relu', padding='same')(x)
12 x = UpSampling2D((2, 2))(x)
13 x = Conv2D(16, (3, 3), activation='relu')(x)
14 x = UpSampling2D((2, 2))(x)
15 decoded = Conv2D(1, (3, 3), activation='sigmoid', padding='same')(x)
16
17 autoencoder = Model(input_img, decoded)
18 autoencoder.compile(optimizer='adam', loss='mse')
```

Autoencoder Architecture

autoencoder.summary()

| Layer (type) | Output Shape | Param # |
|-------------------------------|--------------------|---------|
| input_1 (InputLayer) | (None, 28, 28, 1) | 0 |
| conv2d_1 (Conv2D) | (None, 28, 28, 16) | 160 |
| max_pooling2d_1 (MaxPooling2) | (None, 14, 14, 16) | 0 |
| conv2d_2 (Conv2D) | (None, 14, 14, 8) | 1160 |
| max_pooling2d_2 (MaxPooling2) | (None, 7, 7, 8) | 0 |
| conv2d_3 (Conv2D) | (None, 7, 7, 8) | 584 |
| encoder (MaxPooling2D) | (None, 4, 4, 8) | 0 |
| conv2d_4 (Conv2D) | (None, 4, 4, 8) | 584 |
| up_sampling2d_1 (UpSampling2) | (None, 8, 8, 8) | 0 |
| conv2d_5 (Conv2D) | (None, 8, 8, 8) | 584 |
| up_sampling2d_2 (UpSampling2) | (None, 16, 16, 8) | 0 |
| conv2d_6 (Conv2D) | (None, 14, 14, 16) | 1168 |
| up_sampling2d_3 (UpSampling2) | (None, 28, 28, 16) | 0 |
| conv2d_7 (Conv2D) | (None, 28, 28, 1) | 145 |
| Total params: 4,385 | | |
| Trainable params: 4,385 | | |
| Non-trainable params: 0 | | |

Model Summary

Workflow of the application

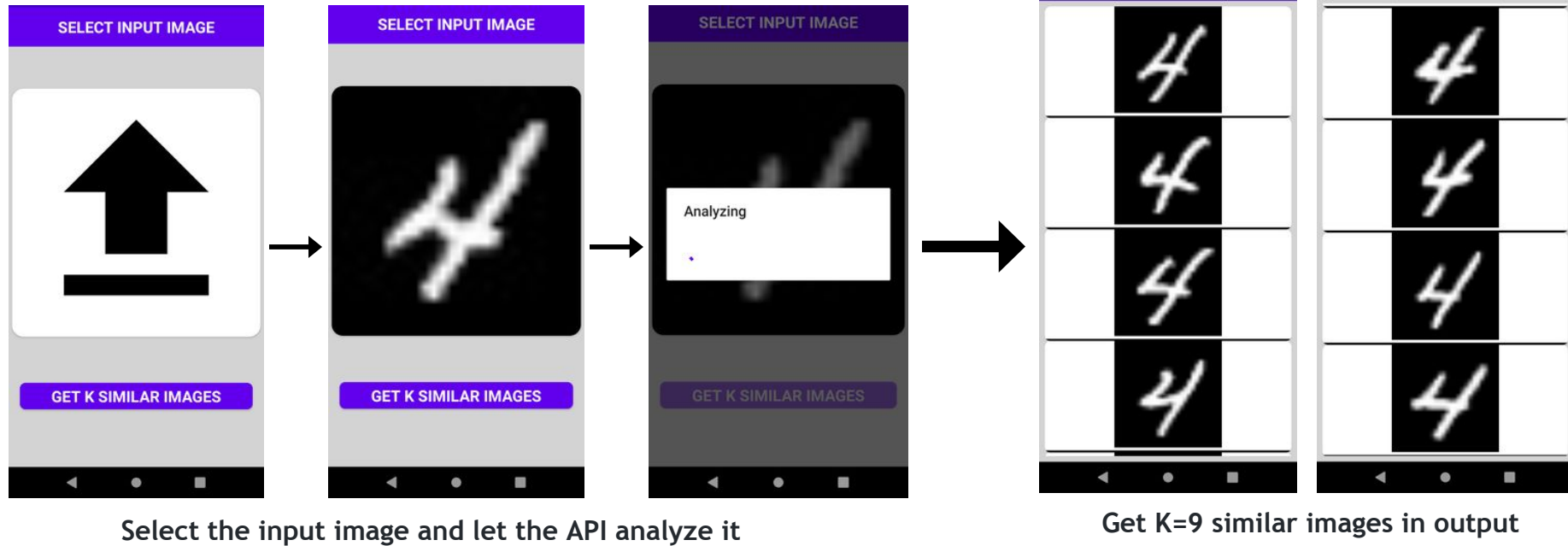
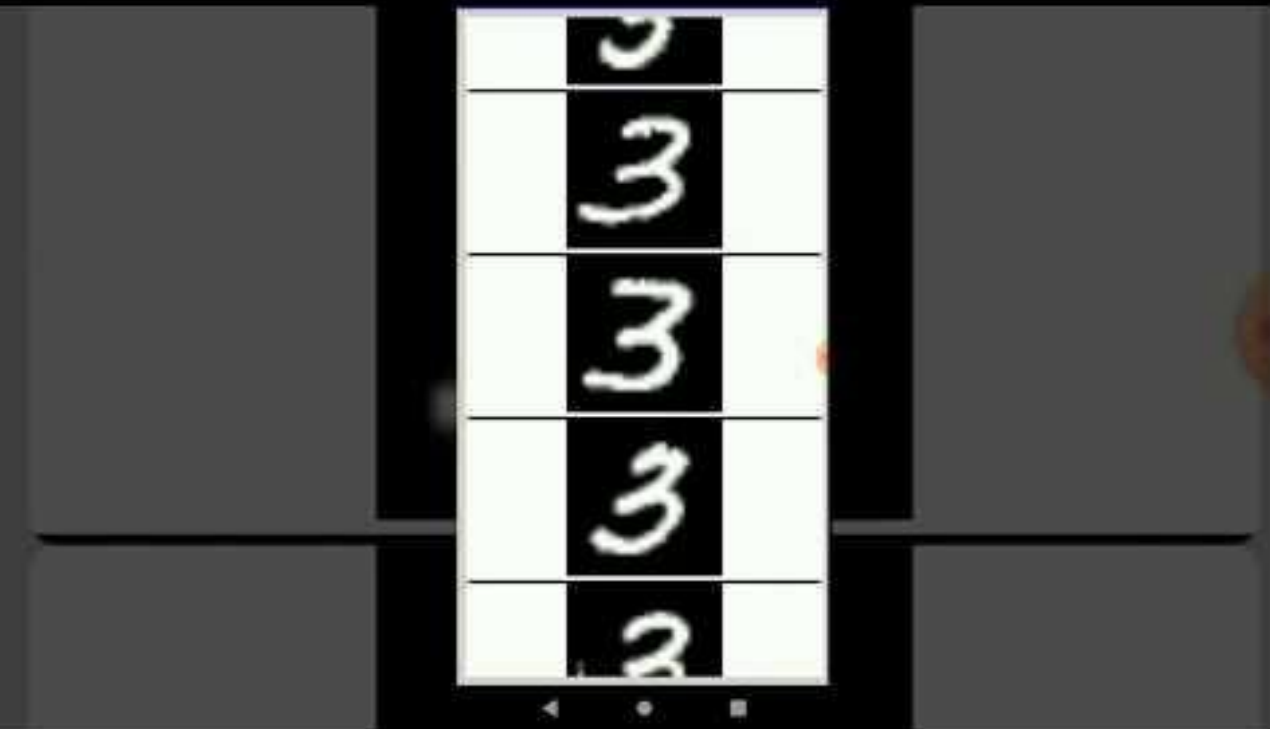


Fig. Screenshots of the Android Application

Demonstration Video & Tech Stack



TECH STACK

Languages :

- Java
- JavaScript
- Python

Tools :

- Visual Studio Code
- Jupyter Notebook
- Android Studio
- Git
- Node.js
- Heroku

Frameworks :

- Express.js
- Tensorflow

Libraries :

- Scikit-Learn
- Keras
- OkHttp
- Glide
- Formidable
- Material.io

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