Automated Personal Writing Pattern Replicator

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Presentation Outline

- Introduction
- Motivation
- Objective
- Scope
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Introduction

- This project focuses on accurately cloning individual handwriting styles using advanced machine learning, enabling the generation of personalized handwritten text and digital fonts.
- It aims to preserve unique handwriting characteristics, to enhance digital personalization and accessibility for various applications, including custom fonts and educational materials.

Motivation

- Preserve unique handwriting styles affected by physical limitations
- Applications in customized fonts and educational materials.
- Aims to address limitations of traditional handwriting cloning methods by ensuring style consistency and handling arbitrary text lengths.

Objective

- Develop a machine learning model to replicate handwriting styles
- Generate consistent, high-fidelity handwritten font for various applications.

Scope

Personalized digital communication

Educational material creation

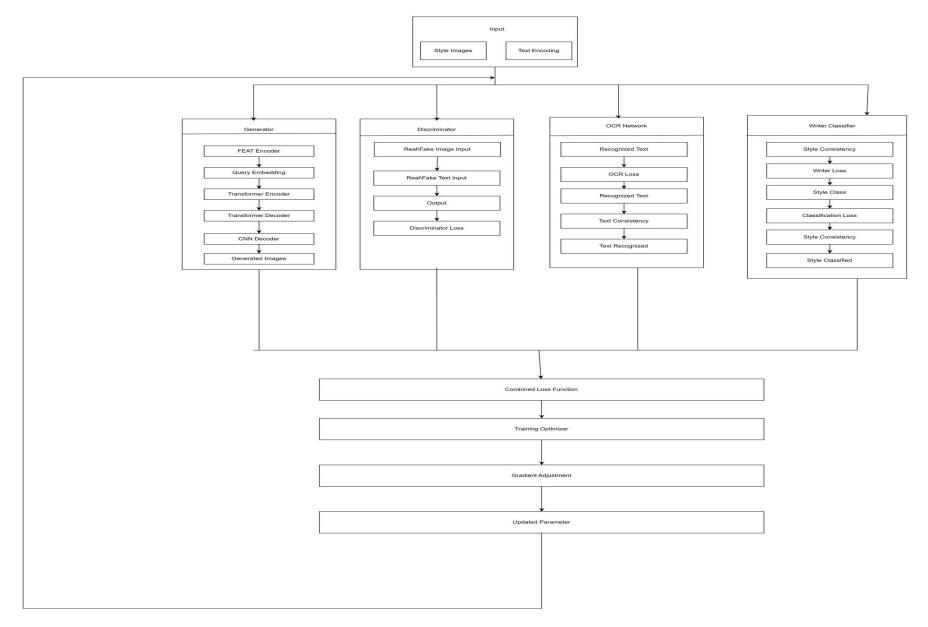
Historical document preservation

Branding and Marketing

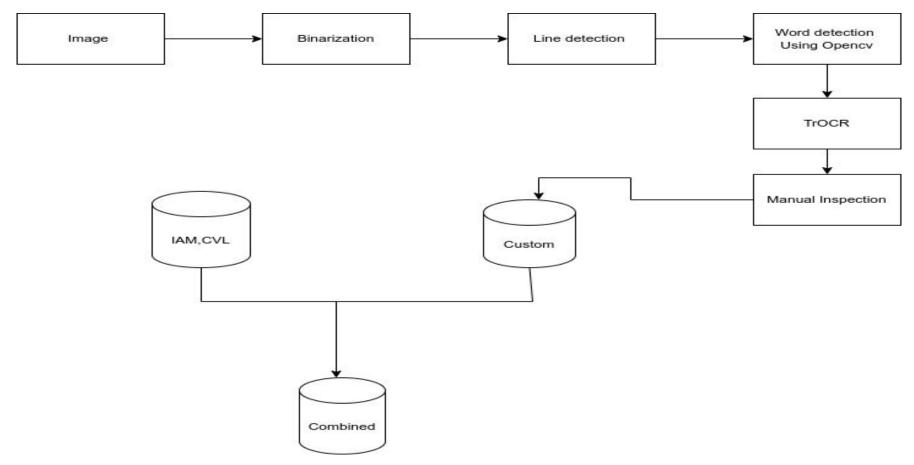
Project Application

- E-commerce Personalization
- Used in Calligraphy
- Custom Handwritten Digital Interfaces
- Personal Blogging and Digital Journals

Methodology(System Diagram)



Methodology (Dataset Preparation)



Methodology(Datasets)

S.N.	Datasets	Writers
1.	IAM Handwriting Dataset	657
2.	CVL Handwriting Dataset	350
3.	Custom Dataset	38

Though they may gather some Leftwing support, a large majority of Labour OM Ps are likely to turn down the Foot-Griffths resolution. Mr. Foot's line will be that as Labour OM Ps opposed the Government Bill which brought life prees into existence, they should not now put Joward nominees. He believes that the House of Roids should be abolished and that Rabour should not take any steps which would appear to "propup" an out-dated institution.

Imagine a vast sheet of paper on which straight Lines, Triangles. Squares, Pentagons, Hexagons, and other figures, instead of remaining fixed in their places, move freely about, on or in the surface, but without the power of rising above or sinking below it, very much like shadows - only hard and with luminous edges - and you will then have a pretty correct notion of my country and countrymen. Alas, a few years ago, I should have said "my universe": but now my mind has been opened to higher views of things.

Imagine a vort sheep of paper on which strought Lines, Triangles, Equares, Penlagons, Hexagons, and other figures, inclead of remaining fixed in their places, more freely about, on or in the surface,

Micro kernels: Micro kernel incorporates only the essential set of OS services into the kernel The rest of the OS services are implemented in program known as 'screen' which aims in user space. This provides a highly modular design and OS-neutral abstraction to the kernel Memory monagement process, times systems and interrupt handless are the essential sources. Examples: Mach, BNX This architecture coters to the problem of ever growing size & kernel code which we could not control in the monolitric approach. This architecture allows some basic services like device triver monagement, protocol stack, file system to run in user space This reduces the kernel code size and also mineases the security and stability of OS as we have the base minimum code summing in the kernel so if suppose a basic service like network service

DENF with Suitable diagram. It is a step the normalization process for database ensuring that all constraints on a relation are expressed as logical implications on the attributes that relation. DKNF requires that the database contains no constraints other than domain constraints and Key constraints. Assume that the domain for wealthy person consists of the name of all wealthy people in a predefined sample of wealthy people The domain for wealthy person for wealthy person type consists of the value "Millionaire" and Billionaire' and the domain for net worth

Fig: Custom Dataset Sample

Methodology (Data Processing)

- Dataset present in IAM , CVL database consists of handwritten pages but for our model training we need to crop out the lines.
- We binarize the image to enhance the quality of the images.
- We extracted the lines by defining the bounding box.
- Further, bounding boxes were defined on the extracted lines to isolate the words present in those lines.

Methodology (Data Processing)[Cont.]

- Passed the extracted words images to TrOCR model to get the transcription.
- The words images were converted to grayscale to simplify processing and reduce computational load.
- Resized images to a consistent size of height 32 pixel and maximum width of 192 pixel to ensure uniformity in the dataset.

relation. DKNF requires that the database contains to constraints other than domain constraints and key constraints. Assume that the domain for wealthy person consists of the name of all wealthy people in a predefined sample of wealthy people



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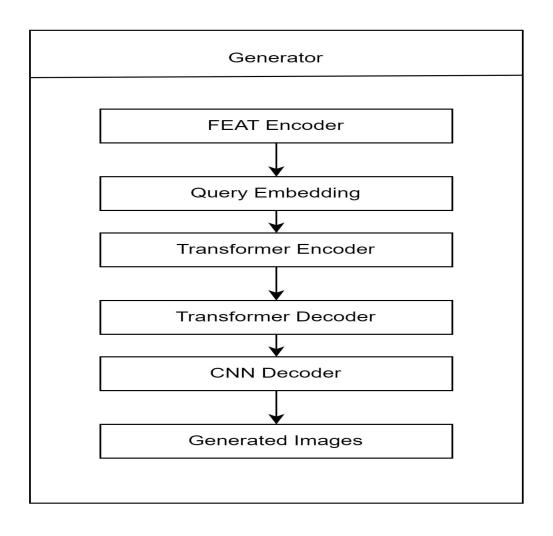


Fig: Custom Dataset Preprocessing

Dataset Overview

<pre>image image - width (px)</pre>	label string · lengths	writer string · classes
16 605	0 45	643 values
in	in	245
all	all	245
parts	parts	245
of	of	245
the	the	245

- **Image**: Contains images of individual handwritten words.
- Label: Includes the text transcription of each handwritten word image.
- Writer: Denotes the writer of each word, aiding in style and writer discrimination.



Methodology (Proposed Model Architecture)-Generator

Style Encoding:

 Feature Encoder: CNN-based network extracts stylistic features from input handwriting examples, converting images into feature representations. The obtained features are flattened and passed to the transformer-based encoder layer.

• **Transformer Encoder**: Processes the style feature sequence to capture handwriting nuances, derived from the CNN based feature encoder.

Methodology (Proposed Model Architecture)-Generator

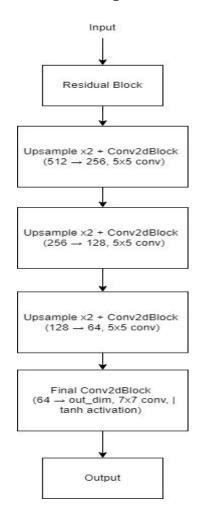
Content Encoding:

• Query Embedding: Transforms the words from the query into embeddings, representing the content that needs to be handwritten.

Decoding:

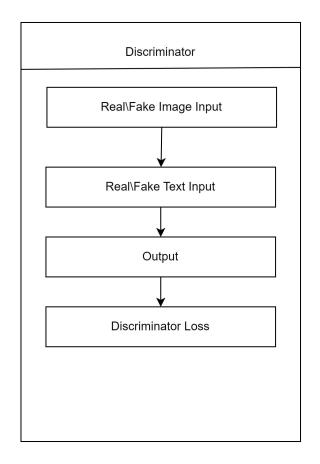
 Transformer Decoder: Merges the style features and the content embeddings to create a unified feature representation to ensures that the generated handwriting maintains the desired style and accurately reflects the content of the query words.

 CNN Decoder: Converts the combined feature representation into the final image of the handwritten text.



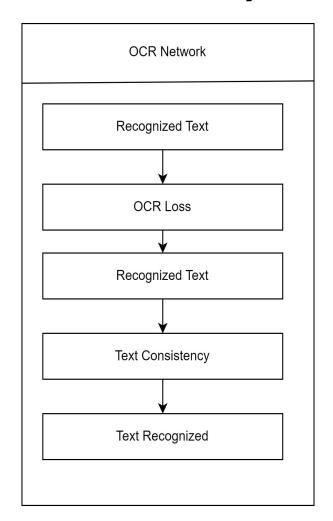
Discriminator:

 Differentiates between real and generated handwriting samples to improve the generator's output through adversarial training.



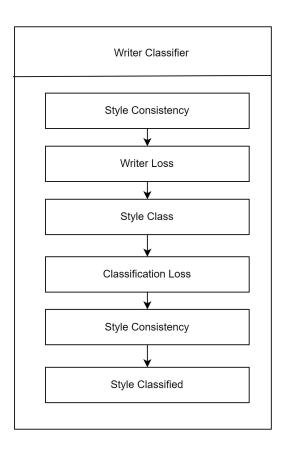
Recognizer:

 Ensures the generated handwriting is legible and correctly represents the input text by recognizing and validating the characters.



Style Classifier:

 Classifies and extracts individual handwriting styles, ensuring the generated handwriting matches the original style attributes of the writer.



Methodology (Evaluation)

FID(Fréchet Inception Distance)

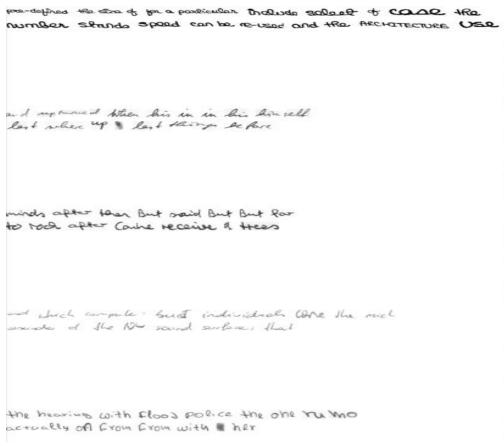
 It is a metric used to measure the similarity between two sets of images

Human Evaluation

 Finally, We perform our model performance using human experience.

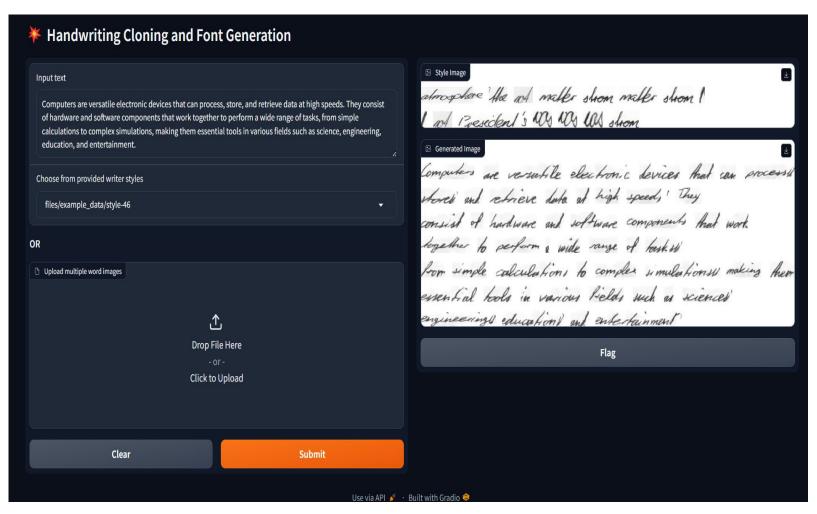
Results & Discussion

 So far we have trained our model on custom dataset that we have prepared and we ran around 963 epochs of training on that dataset and saw the result in validation set like this:

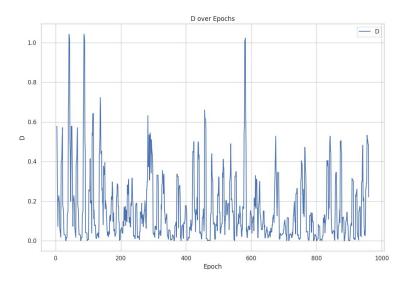


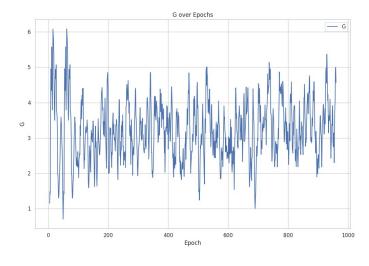
The Statue of Liberty's arguably one of New book city is most iconic symbolist is a popula tourist attraction for first time vieritors This tob-foot monument was gifted to the on States from France in order to celebrate You years of America's independence When Claire Statue of Liberty for the first times she instantly admired it as a symbol of freedom The Statue of ciserty arguably one of Hew Jork City's most iconic symbolso is a signiti townist attraction for first - time visitors This Fil - food unument was gifted to the Un States from France in order to celebrate PC years of America's independence when Claire Stake of Liberty for the first times she instantly admired it as a symbol of freedom The Status of Liberty's arguably one of Hew book city's most icomic symbolso is a popula tourist attraction for first - time visitors t This Hel-Post monument was gifted to the Un States from France in order to celebrate 466 years of America's independence When Claire Status of Liberty for the first times she instantly admired it as a symbol of freedom The Statue of Liberty warquably one of New fork City's most iconic symbolio is a popul tourist attraction for first - time visition. This DD - food monument was gifted to the Un States from Trance in order to calabrate Ville year of America's independence When Claire Statue of Liberty for the first times she instantly admired it as a symbol of breedom The Status of Libertyo arguebly one of Hew York City's Most iconic symboleo is a popula tourist attraction for first time visitors !

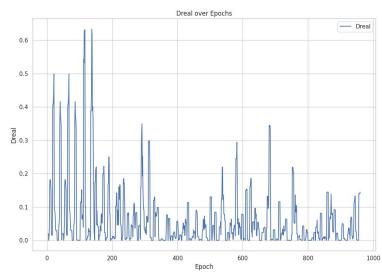
Result

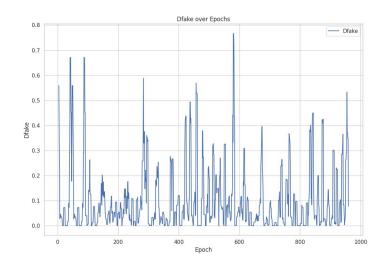


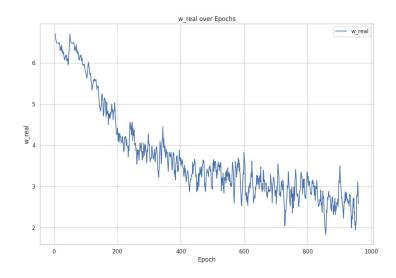
Loss

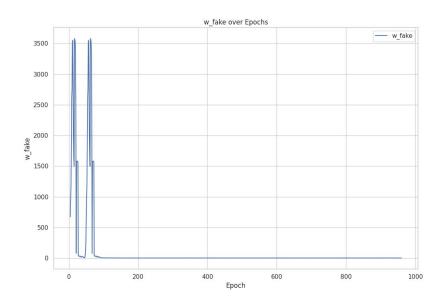


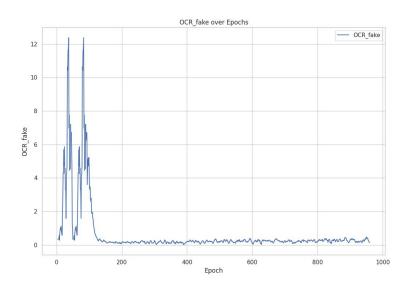


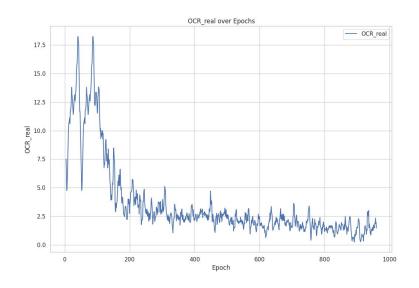












Loss Analysis

- **loss-OCR_real**: 0.9483
 - OCR network loss on real data measures the OCR network's ability to recognize real handwriting.
- loss-w_fake: 1.0213
 - Writer classification loss on fake data indicates how well the classifier identifies the writer of generated handwriting.
- loss-w_real: 2.3178
 - Writer classification loss on real data indicates how well the classifier identifies the writer of real handwriting.

Loss Analysis

- **loss-G**: 4.4243
 - Generator loss (G) measures how well the generator produces realistic data.
- **loss-D**: 0.0215
 - Overall Discriminator loss (D) measures the discriminator's ability to differentiate between real and fake data.
- **loss-D Fake**: 0.0
 - Discriminator loss on fake data (Dfake) indicates how well the discriminator identifies generated data as fake.
- **loss-Dreal**: 0.0215
 - Discriminator loss on real data (Dreal) shows how well the discriminator identifies real data as real.
- **loss-OCR_fake**: 0.1757
 - OCR network loss on fake data measures the OCR network's ability to recognize generated handwriting.

Evaluation

While Evaluating the generated handwriting FID Score was found to be 112.55345241037465

Remaining Task

- To further train the model and check model accuracy.
- To diversify our dataset including alphanumeric dataset.
- To work on creating font of generated dataset.

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