



# Synthetic Image Generation

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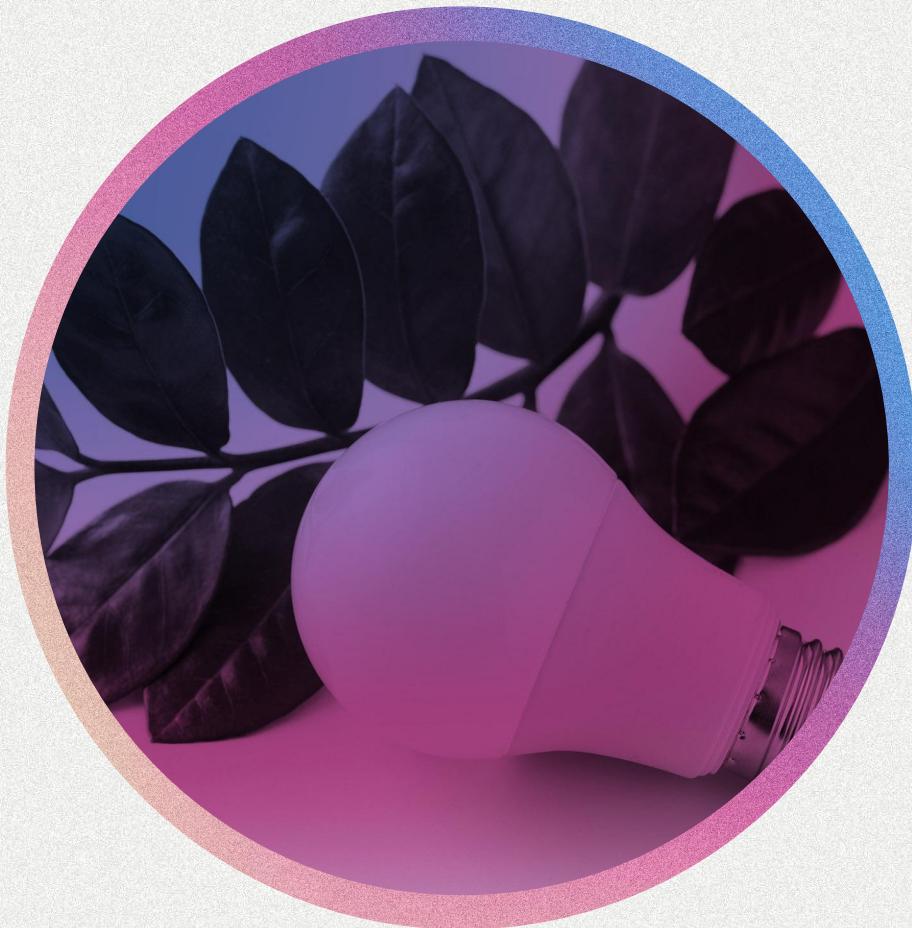
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# **Idea, Motivation & How it started**





## Idea & Motivation

### 1. Bridging Text and Image:

- Develop a generative model that translates textual descriptions/prompts into lifelike flower images.

### 2. User-Friendly Platform:

- Create a platform that enables users to express their thoughts in the form of flower images effortlessly.

### 3. Empower Verbally Impaired Individuals

- Enable and enhance the creativity, independence, and self-expression

# Business Actions

01

## Platform Development

- Prioritize ease of use, intuitive interfaces, and accessibility features to ensure that visually impaired individuals can easily navigate and utilize the system.

02

## User Engagement

- Reach out to differently abled individuals and organizations
- Conduct awareness campaigns and organize workshops
- Collaborate with disability advocates and influencers

03

## User Support

- Comprehensive user support and training resources
- Prompt and responsive customer support channels

# How it started



## Dataset Selection

Selecting a flower dataset that contains 6552 images and captions of different varieties of flowers.

## Training GANs Model

Building a stacked GANs Model from scratch proved to be very computationally heavy

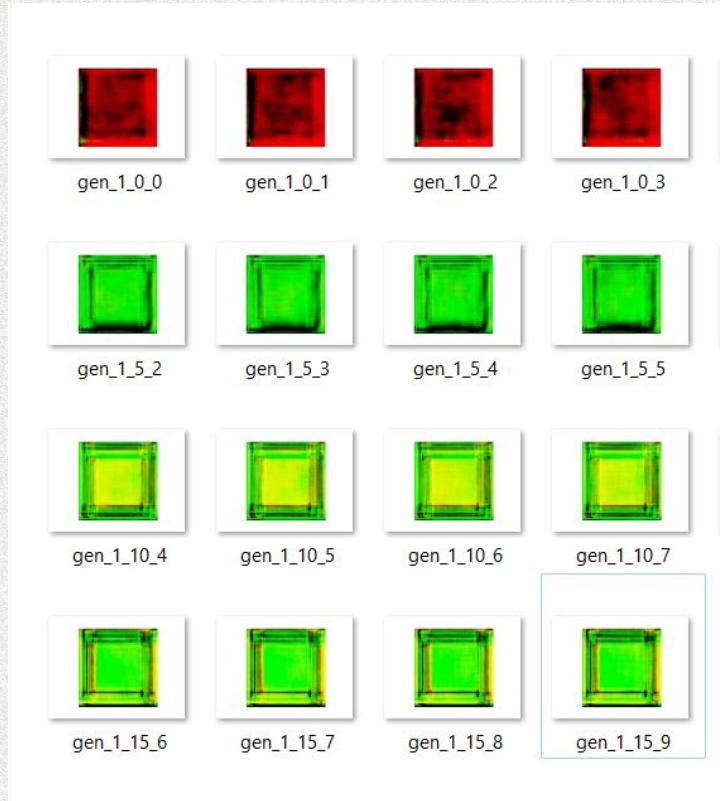
## Fine-tuning Stable Diffusion Model

Applying a pre-trained text and image encoders to our diffusion model which provided us the tuned parameters.

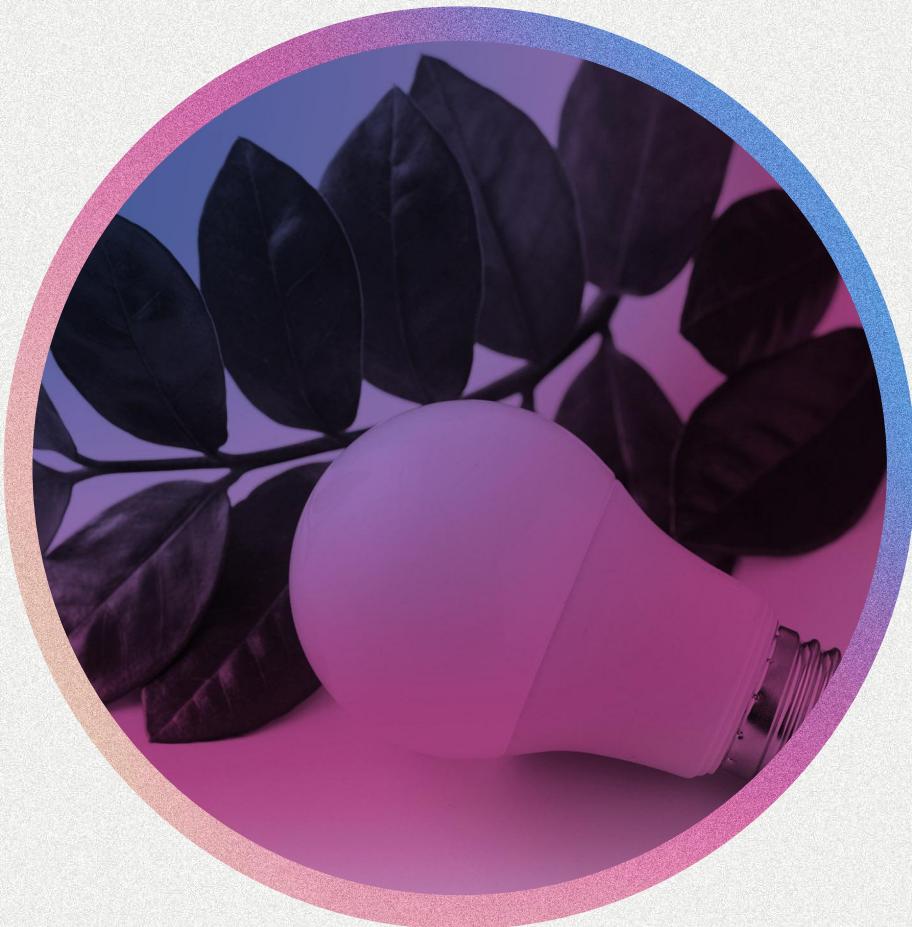
## Loading Weights

Loading and saving the weights generated from our model to be used later for new flower image generation

# Results of the Trained GANs Model



# **Methodology & Output**



# Literature Behind Diffusion Models

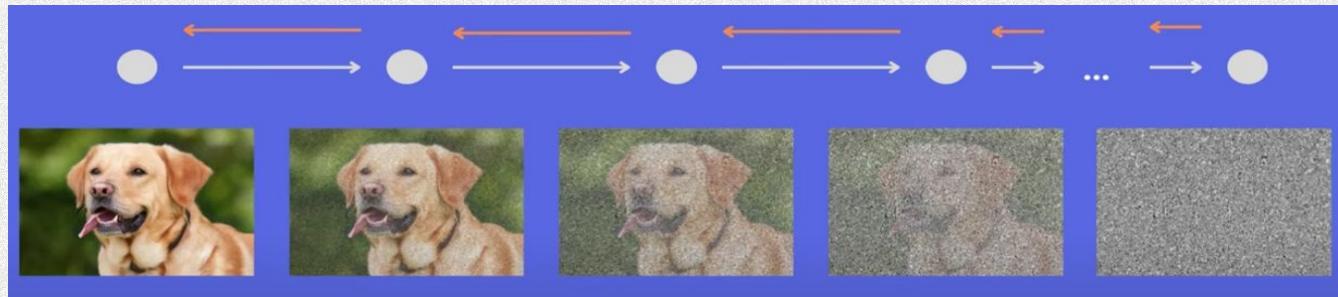
**Diffusion models were inspired from non-equilibrium thermodynamics from physics.**



- With diffusion models the goal is to learn a model that can reverse the diffusion process and bring the drop of paint to its original state.
- The drop being in one spot includes some information and as the diffusion process progresses, we lose this information.
- The one drop being the clear images we're interested in.

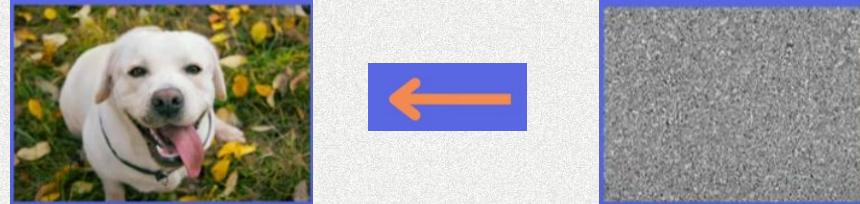
# How do Diffusion models work?

**Diffusion models work by adding noise to original images and later learning how to reverse this noise process through a markov chain**



In each time step we are adding a little bit of noise to our image until the image only consist of noise and later learning how to reverse this noise adding process.

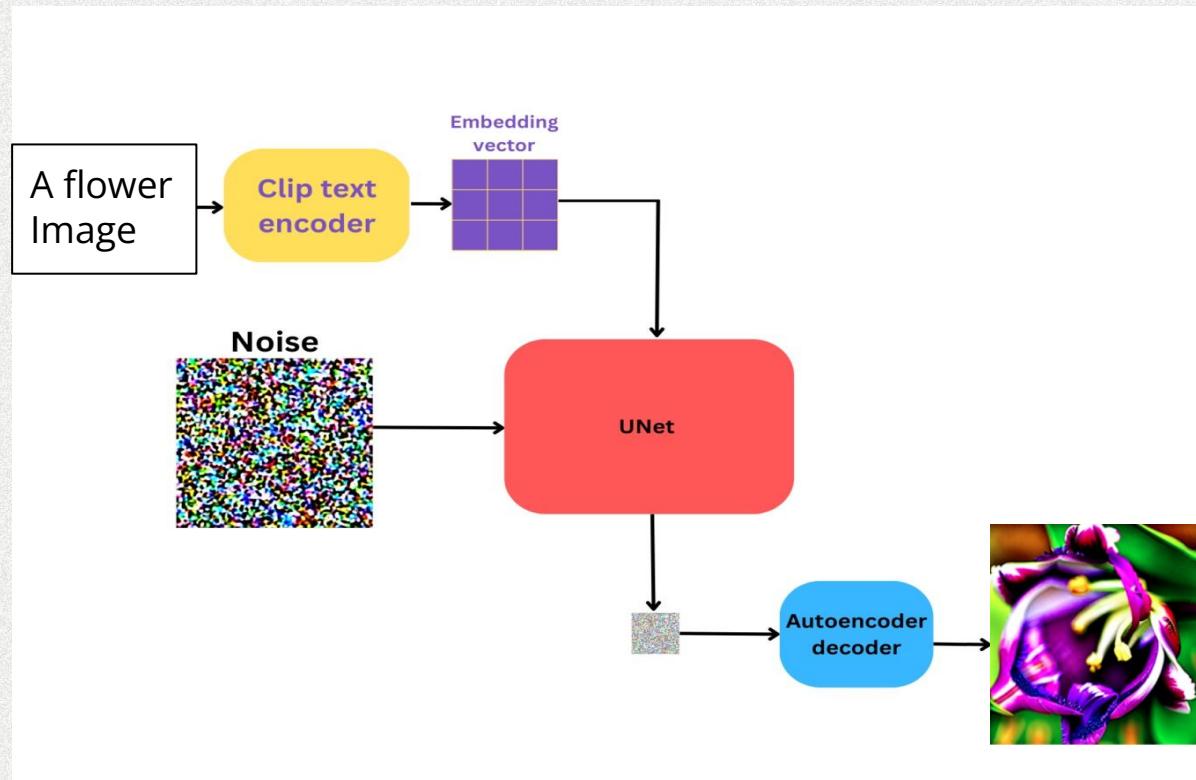
**After the model is trained given only noise, the model is able to generate high resolution images.**



# **What we did**



# 1. Model



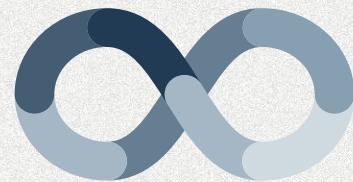
## 2. Data Preparation

- Selected a Pre-trained Diffusion Model:
- Pre-processed the data to match the input format expected by the diffusion model
- Total Images: 6550
- Textual Descriptions: ~9/image
- Tokenized the textual descriptions.
- Processed the images.
- Encoded the tokenized texts.
- Prepared the training dataset.

image (image)	text (string)
	"a close up of a passion flower"
	"a blue passion flower with a yellow flower in the middle of the flower"
	"a purple passion flower in a pot"
	"a purple passion flower with white flowers and green leaves"

### 3. Model Parameters

- Image Encoder, Text Encoder Form the original Pre-trained model
- Maximum length for a text prompt: 77
- Image Resolution: 256
- Generated Normal Random Noise
- Hyperparameters: [Tutorial from “Hugging Face”](#)
- Loss: 0.2285



## 4.Image Generation and Evaluation



- Loaded the weights of the fine-tuned diffusion model from the saved file.
- Generated flower images based on different prompts using the `text_to_image` function of the Stable Diffusion Model.
- Plotted and visualized the generated images along with their corresponding prompts.

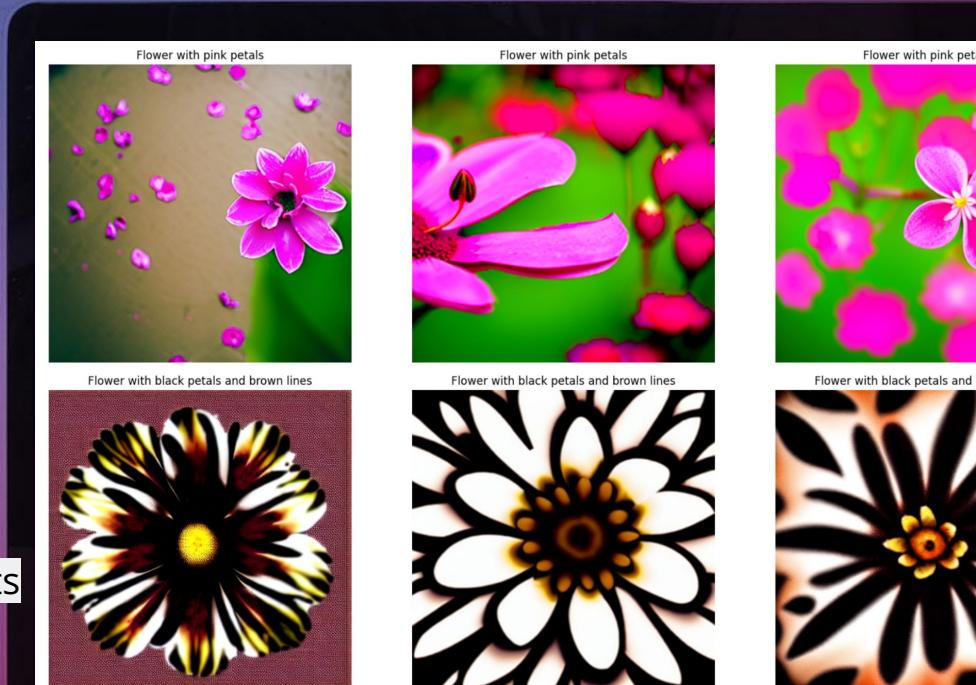
# Model Output

## Textual Prompts:

- “Flowers with pink petals”
- “Flowers with black petals and brown lines”

## Image Output:

- Visually coherent and representative
- Strong alignment between the textual prompts and the generated flower images
- Produces a diverse range of flower images





**Thank  
You!**