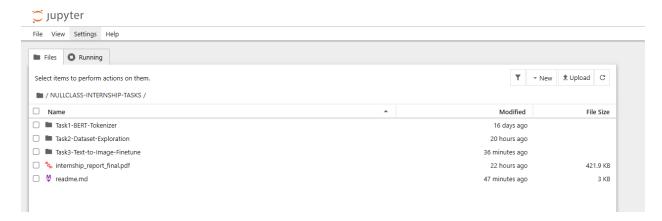
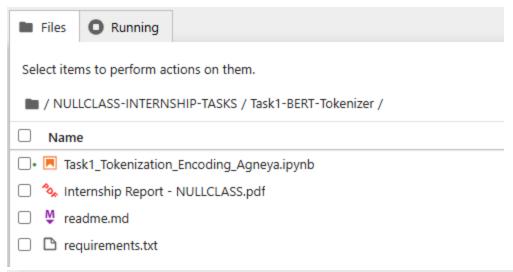
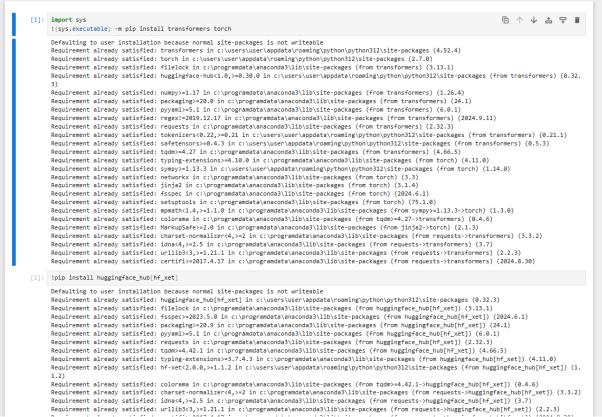
SCREENSHOTS

AUTHOR: AGNEYAS NAMBIAR

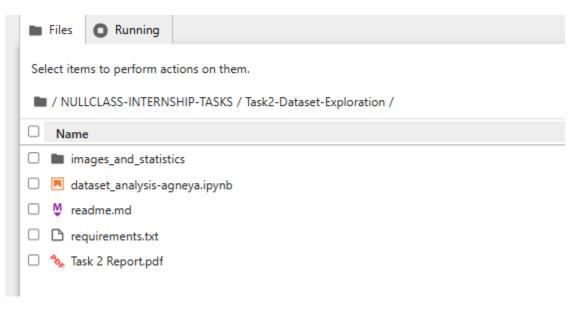




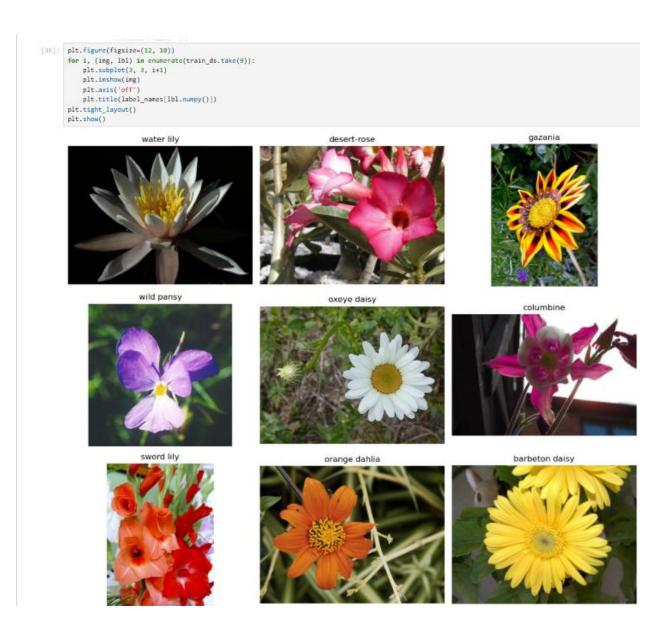


```
[11]: from transformers import BertTokenizer, BertModel
      import torch
      import numpy
      import pandas
[12]: tokenizer = BertTokenizer.from_pretrained('bert-base-uncased')
      model = BertModel.from_pretrained('bert-base-uncased')
[13]: text = "Hello, my name is agneya and i am doing internship with nullclass"
[14]: #Step 1: Tokenization (splitting into tokens)
      tokens = tokenizer.tokenize(text)
      print("Tokens:")
      print(tokens)
      ['hello', ',', 'my', 'name', 'is', 'ag', '##ney', '##a', 'and', 'i', 'am', 'doing', 'internship', 'with', 'null', '##class']
[15]: #Step 2: Encoding (tokens → input IDs + tensors)
      encoding = tokenizer(text, return_tensors='pt')
[16]: print("Input IDs:")
      print(encoding['input_ids'])
      Input IDs:
      tensor([[ 101, 7592, 1010, 2026, 2171, 2003, 12943, 5420, 2050, 1998, 1045, 2572, 2725, 22676, 2007, 19701, 26266, 102]])
[17]: print("Attention Mask:")
      print(encoding['attention_mask'])
      Attention Mask:
      [18]: #Step 3:Use BERT to get embeddings
      with torch.no_grad():
         outputs = model(**encoding)
      print("\nOutput Embeddings Shape (last_hidden_state):")
      print(outputs.last_hidden_state.shape)
      Output Embeddings Shape (last_hidden_state):
      torch.Size([1, 18, 768])
[19]: #tokenizer.tokenize()=Splits sentence into wordpieces
      #tokenizer(text, return_tensors='pt')=Converts text into token IDs and attention mask
      #model(**encoding)=Feeds input to BERT to get embeddings
      #outputs.last_hidden_state.shape=Returns [1, sequence_length, 768] - the BERT embeddings
```

TASK 2



```
[2]: pip install tensorflow-datasets tensorflow matplotlib pandas numpy
                                                                                                                                          □ ↑ ↓ 占 〒 ■
         Defaulting to user installation because normal site-packages is not writeable
         Requirement already satisfied: tensorflow-datasets in c:\users\user\appdata\roaming\python\python312\site-packages (4.9.9)
Requirement already satisfied: tensorflow in c:\users\user\appdata\roaming\python\python312\site-packages (2.19.0)
         Requirement already satisfied: matplotlib in c:\programdata\anaconda3\lib\site-packages (3.9.2)
         Requirement already satisfied: pandas in c:\programdata\anaconda3\lib\site-packages (2.2.2)
         Requirement already satisfied: nummy in c:\programdata\anaconda3\lib\site-packages (1.26.4)
Requirement already satisfied: abs1-py in c:\users\user\appdata\roaming\python\python312\site-packages (from tensorflow-datasets) (2.3.0)
         Requirement already satisfied: dm-tree in c:\users\user\appdata\roaming\python\python312\site-packages (from tensorflow-datasets) (0.1.9)
         Requirement already satisfied: etils>=1.9.1 in c:\users\user\appdata\roaming\python\python312\site-packages (from etils[edc,enp,epath,epy,etree]>=1.
         9.1; python_version >= "3.11"->tensorflow-datasets) (1.12.2)
         Requirement already satisfied: immutabledict in c:\users\user\appdata\roaming\python\python312\site-packages (from tensorflow-datasets) (4.2.1)
         Requirement already satisfied: promise in c:\users\user\appdata\roaming\python\python312\site-packages (from tensorflow-datasets) (2.3)
         Requirement already satisfied: protobuf>=3.20 in c:\programdata\anaconda3\lib\site-packages (from tensorflow-datasets) (4.25.3) Requirement already satisfied: psutil in c:\programdata\anaconda3\lib\site-packages (from tensorflow-datasets) (5.9.0)
         Requirement already satisfied: pyarrow in c:\programdata\anaconda3\lib\site-packages (from tensorflow-datasets) (16.1.0)
Requirement already satisfied: requests>=2.19.0 in c:\programdata\anaconda3\lib\site-packages (from tensorflow-datasets) (2.32.3)
         Requirement already satisfied: simple_parsing in c:\users\user\appdata\roaming\python\python312\site-packages (from tensorflow-datasets) (0.1.7)
         Requirement already satisfied: tensorflow-metadata in c:\users\user\appdata\roaming\python\python312\site-packages (from tensorflow-datasets) (1.17. 🔀
    [3]: import tensorflow_datasets as tfds
         import matplotlib.pyplot as plt
         import pandas as pd
         import numpy as np
         import os
         from pathlib import Path
   [4]: tfds.disable progress bar()
   [7]: dataset, info = tfds.load('oxford_flowers102', with_info=True, as_supervised=True)
         train_ds, val_ds, test_ds = dataset['train'], dataset['validation'], dataset['test']
   [9]: print("Number of classes:", info.features['label'].num_classes)
         Number of classes: 102
[10]: print("Total examples:")
         print(" - Train:", info.splits['train'].num_examples)
         print(" - Validation:", info.splits['validation'].num_examples)
         print(" - Test:", info.splits['test'].num_examples)
         Total examples:
          - Train: 1020
          - Validation: 1020
          - Test: 6149
[11]: label_names = info.features['label'].names
[49]: label_names = info.features['label'].names
[14]: res_df = pd.DataFrame(image_shapes, columns=['Height', 'Width', 'Channels'])
         print("Image resolution statistics (100 samples):")
         print(res_df.describe())
         Image resolution statistics (100 samples):
                                     Width Channels
                      Height
         count 100.000000 100.000000
                                                     100.0
         mean 541.670000 623.080000
                                                         3.0
         std
                   78.265806 106.863046
                                                          0.0
         min
                  500.000000 500.000000
                                                          3.0
         25%
                  500.000000 500.000000
                                                          3.0
         50%
                 500.000000 660.500000
                                                         3.0
         75%
                 544.250000 726.750000
                                                         3.0
                 883.000000 825.000000
                                                          3.0
[16]: print(f"Number of classes: {info.features['label'].num_classes}")
         Number of classes: 102
```



Flower Class: 72



```
[35]: os.makedirs("images_and_statistics", exist_ok=True)
dataset = tfds.load('oxford_flowers102', split='train', as_supervised=True)
plt.figure(figsize=(10, 10))
for i, (image, label) in enumerate(dataset.take(9)):
    plt.subplot(3, 3, i + 1)
    plt.sinbow(image.numpy())
    plt.title(f"class ID: {label.numpy()}")
              plt.axis("off")
plt.tight_layout()
plt.savefig("images_and_statistics/flower_grid.png")
plt.show()
                                                                                                                                                                                                                               Class ID: 70
                                            Class ID: 72
                                                                                                                                      Class ID: 84
                                             Class ID: 51
                                                                                                                                      Class ID: 48
                                                                                                                                                                                                                               Class ID: 83
                                             Class ID: 42
                                                                                                                                      Class ID: 58
                                                                                                                                                                                                                                Class ID: 40
```

```
[44]: output_dir = "images_and_statistics"
           os.makedirs(output_dir, exist_ok=True)
dataset = tfds.load("oxford_flowers102", split="train", as_supervised=True)
            image_sizes = []
            for image, label in dataset.take(200): # Analyze first 200 images
h, w = image.shape[0], image.shape[1]
image_sizes.append((h, w))
[45]: df_sizes = pd.DataFrame(image_sizes, columns=["Height", "width"]) csv_path = os.path.join(output_dir, "image_size_stats.csv") df_sizes.to_csv(csv_path, index=False)
[46]: plt.figure(figsize=(8, 6))
    df_sizes.boxplot()
    plt.title("Image Size Distribution")
    plt.ylabel("Pixels")
           ptt.grid(True)
boxplot_path = os.path.join(output_dir, "image_size_boxplot.png")
plt.savefig(boxplot_path)
           plt.show()
                                                                         Image Size Distribution
                900
                                                           0
                                                           0
                850
                800
                750
                                                           0
            Pixels 200
                                                           8
                650
                600
                550
                500
```

Width

Height

[47]: plt.figure(figsize=(10, 10))
for i, (image, label) in enumerate(dataset.take(9)):
 plt.subplot(3, 3, i + 1)
 plt.imshow(image.numpy())
 plt.title(f"class ID: {label.numpy()}")
 plt.axis("off")











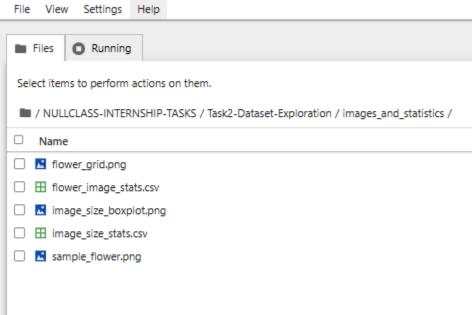


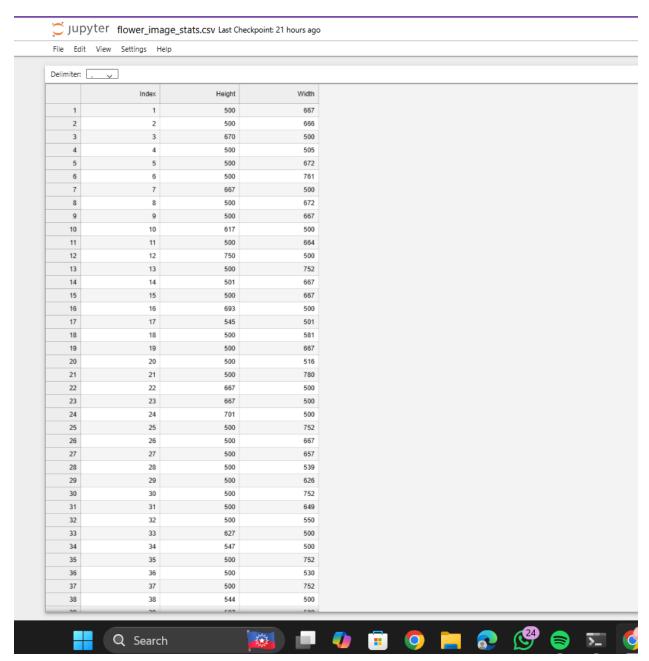






```
[1]: import tensorflow_datasets as tfds
       import pandas as pd
      # Load dataset (as supervised gives (image, Label) tuple)
ds, ds_info = tfds.load("oxford_flowers102", split='train', as_supervised=True, with_info=True)
       # Sample 200 images and collect dimensions
       image_sizes = []
       for i, (image, label) in enumerate(ds.take(200)):
           height = image.shape[0]
            width = image.shape[1]
           image_sizes.append({'Index': i+1, 'Height': height, 'Width': width})
       # Convert to DataFrame
       image_sizes_df = pd.DataFrame(image_sizes)
       image_sizes_df.to_csv('images_and_statistics/flower_image_stats.csv', index=False)
[2]: import seaborn as sns
       import matplotlib.pyplot as plt
      df = pd.read_csv('images_and_statistics/flower_image_stats.csv')
sns.boxplot(data=df[['Height', 'Width']])
plt.title("Image Size Distribution (Height vs Width)")
       \verb"plt.savefig" ("images_and_statistics/image_size_boxplot.png")"
                         Image Size Distribution (Height vs Width)
       900
                                 0
       850
       800
                                 000000
       750
       700
                                 8
       650
       600
       550
       500
                                                                       Width
```





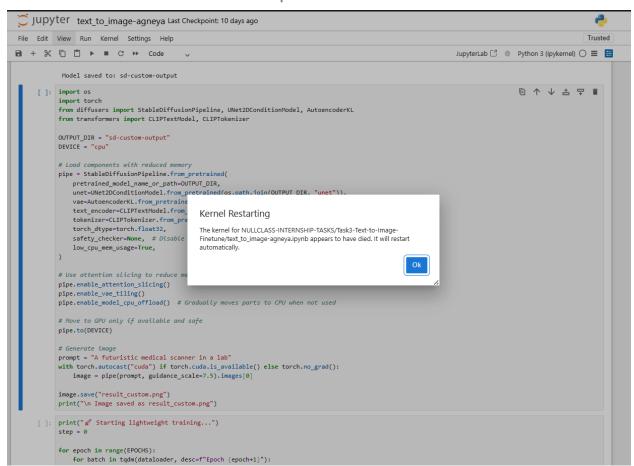
TASK 3

```
[1]: import os
      import csv
      from PIL import Image
[2]: dataset_dir = "dataset"
      images_dir = os.path.join(dataset_dir, "images")
      os.makedirs(images_dir, exist_ok=True)
[3]: img_names = ["img1.jpg", "img2.jpg", "medical_scan_01.jpg"] colors = [(255, 0, 0), (0, 255, 0), (0, 0, 255)] # Red, Green, Blue
      for name, color in zip(img_names, colors):
          img = Image.new("RGB", (512, 512), color)
          img.save(os.path.join(images_dir, name))
[4]: captions = [
          ("img1.jpg", "A red flower in a green field"),
("img2.jpg", "A futuristic medical scanner in a lab"),
          ("medical_scan_01.jpg", "High-resolution MRI scan showing brain activity"),
      csv_path = os.path.join(dataset_dir, "captions.csv")
      with open(csv_path, "w", newline="") as f:
         writer = csv.writer(f)
         writer.writerow(["file_name", "text"])
          writer.writerows(captions)
      print("Dataset prepared!")
      Dataset prepared!
[5]: import torch
      import numpy as np
      import pandas as pd
[6]: from torch.utils.data import Dataset, DataLoader
      from transformers import CLIPTokenizer, CLIPTextModel
      from diffusers import AutoencoderKL, UNet2DConditionModel, StableDiffusionPipeline, DDIMScheduler
[7]: from accelerate import Accelerator
      from tqdm import tqdm
```

```
[8]: class CustomImageDataset(Dataset):
    def __init__(self, images_dir, captions_file, tokenizer, size=512):
                self.images_dir = images_dir
                self.data = pd.read_csv(captions_file)
self.tokenizer = tokenizer
                 self.size = size
           def len (self):
                 return len(self.data)
            def __getitem__(self, idx):
                 row = self.data.iloc[idx]
                image_path = os.path.join(self.images_dir, row['file_name'])
image = Image.open(image_path).convert("RGB").resize((self.size, self.size))
                 image = torch.tensor(np.array(image)).permute(2, 0, 1).float() / 255.0
                inputs = self.tokenizer(
                     truncation=True.
                     padding="max_length",
                     max_length=self.tokenizer.model_max_length,
                     return_tensors="pt"
                     "pixel_values": image,
                     "input_ids": inputs.input_ids.squeeze(0)
[9]: PRETRAINED_MODEL_NAME = "CompVis/stable-diffusion-v1-4"
       DEVICE = "cuda" if torch.cuda.is_available() else "cpu"
       SEED = 42
       torch.manual seed(SEED)
[9]: <torch._C.Generator at 0x18553fffb90>
[10]: tokenizer = CLIPTokenizer.from_pretrained(PRETRAINED_MODEL_NAME, subfolder="tokenizer")
       text_encoder = CLIPTextModel.from_pretrained(PRETRAINED_MODEL_NAME, subfolder="text_encoder").to(DEVICE)
vae = AutoencoderKL.from_pretrained(PRETRAINED_MODEL_NAME, subfolder="vae").to(DEVICE)
       unet = UNet2DConditionModel.from_pretrained(PRETRAINED_MODEL_NAME, subfolder="unet").to(DEVICE)
       dataset = CustomImageDataset(
           images_dir=images_dir,
            captions_file=csv_path,
           tokenizer=tokenizer
       dataloader = DataLoader(dataset, batch_size=2, shuffle=True)
[13]: print("Starting training...")
       MAX_TRAIN_STEPS = 10
       for epoch in range(1):
   for batch in tqdm(dataloader, desc=f"Epoch {epoch+1}"):
                pixel_values = batch["pixel_values"].to(DEVICE)
input_ids = batch["input_ids"].to(DEVICE)
                with torch.no_grad():
    encoder_hidden_states = text_encoder(input_ids)[0]
                    latents = vae.encode(pixel_values).latent_dist.sample() * 0.18215
                noise = torch.randn_like(latents)
                timesteps = torch.randint(0, 1000, (latents.shape[0],), device=DEVICE).long()\\
                noisy_latents = noise_scheduler.add_noise(latents, noise, timesteps)
                noise_pred = unet(noisy_latents, timesteps, encoder_hidden_states).sample
loss = torch.nn.functional.mse_loss(noise_pred, noise)
                optimizer.zero_grad()
                accelerator.backward(loss)
                optimizer.step()
                    print(f"Step {step} | Loss: {loss.item():.4f}")
                torch.cuda.empty_cache()
                if step >= MAX_TRAIN_STEPS:
            if step >= MAX_TRAIN_STEPS:
                print("\n Training complete.")
       Starting training...
       Epoch 1: 50%
                                                                                               | 1/2 [04:09<04:09, 249.63s/it]
       Epoch 1: 100%|
                                                                                               2/2 [07:37<00:00, 228.65s/it]
```

```
[14]: # Save model
OUTPUT_DIR = "sd-custom-output"
if accelerator.is_main_process:
    os.makedirs(OUTPUT_DIR, exist_ok=True)
    unet.save_pretrained(os.path.join(OUTPUT_DIR, "unet"))
    vae.save_pretrained(os.path.join(OUTPUT_DIR, "vae"))
    text_encoder.save_pretrained(os.path.join(OUTPUT_DIR, "text_encoder"))
    tokenizer.save_pretrained(os.path.join(OUTPUT_DIR, "tokenizer"))
    print(f"\n Model saved to: {OUTPUT_DIR}")
```

Model saved to: sd-custom-output



Full trained pipeline or sd-custom-output folder is heavy, Jupyter Notebook kernel keeps restarting.

The model or pipeline is too large to load fully into your system's RAM/VRAM.

You're hitting **memory overflow**, especially during:

- StableDiffusionPipeline.from_pretrained(...)
- Running pipe(prompt) for image generation.