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                                                                                                                                                                                                      NLP_TA2_Project.ipynb - Colab

    TA2: Marathi Handwriting/Text Recognition using Transformer-based OCR + NLP

   Steps to Follow:
           1. Capture or upload an image of a handwritten page written in your mother tongue.
           2. Preprocess the image to improve clarity (e.g., convert to grayscale, resize, denoise).
           3. Use OCR to extract text from the image (e.g., Transformer based OCR with language pack for your language).
           4. Normalize the extracted text (remove noise, unwanted characters, fix encoding issues).
           5. Tokenize the text using an appropriate NLP tokenizer for your language.
           6. Marathi Named Entity Recognition (NER), Perform Named Entity Recognition on the extracted Marathi text to identify
             entities such as names of people, places, organizations, dates, etc.
           7. Performing NLP task, such as:

    Language detection

    Translation

           8. Display the final output in a readable format (console, notebook cell, or GUI).
           9. Sentiment Analysis:
             Analyze the sentiment (positive, negative, or neutral) of the text extracted through OCR to understand the emotional tone
             of the content.
          10. Summarization of the extracted text

	✓ Step 0: Install & Import Required Libraries

   # Install All Required Python Libraries
    !pip install gdown
    !pip install indic-nlp-library
    !pip install pytesseract
    !pip install opencv-python-headless
    !pip install googletrans==4.0.0-rc1
    !pip install deep-translator
    !pip install langdetect
   # Update and Install All Required Libraries and Tools
    !sudo apt-get update
    !sudo apt-get upgrade
    !sudo apt-get install -y tesseract-ocr
    !sudo apt-get install -y tesseract-ocr-mar
    !git clone https://github.com/anoopkunchukuttan/indic_nlp_resources.git
    Requirement already satisfied: gdown in /usr/local/lib/python3.11/dist-packages (5.2.0)
        Requirement already satisfied: beautifulsoup4 in /usr/local/lib/python3.11/dist-packages (from gdown) (4.13.3)
        Requirement already satisfied: filelock in /usr/local/lib/python3.11/dist-packages (from gdown) (3.18.0)
        Requirement already satisfied: requests[socks] in /usr/local/lib/python3.11/dist-packages (from gdown) (2.32.3)
        Requirement already satisfied: tqdm in /usr/local/lib/python3.11/dist-packages (from gdown) (4.67.1)
        Requirement already satisfied: soupsieve>1.2 in /usr/local/lib/python3.11/dist-packages (from beautifulsoup4->gdown) (2.6)
        Requirement already satisfied: typing-extensions>=4.0.0 in /usr/local/lib/python3.11/dist-packages (from beautifulsoup4->gdown) (4.13.1)
        Requirement already satisfied: charset-normalizer<4,>=2 in /usr/local/lib/python3.11/dist-packages (from requests[socks]->gdown) (3.4.1)
        Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.11/dist-packages (from requests[socks]->gdown) (3.10)
        Requirement already satisfied: urllib3<3,>=1.21.1 in /usr/local/lib/python3.11/dist-packages (from requests[socks]->gdown) (2.3.0)
        Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.11/dist-packages (from requests[socks]->gdown) (2025.1.31)
        Requirement already satisfied: PySocks!=1.5.7,>=1.5.6 in /usr/local/lib/python3.11/dist-packages (from requests[socks]->gdown) (1.7.1)
        Collecting indic-nlp-library
          Downloading indic_nlp_library-0.92-py3-none-any.whl.metadata (5.7 kB)
        Collecting sphinx-argparse (from indic-nlp-library)
          Downloading sphinx_argparse-0.5.2-py3-none-any.whl.metadata (3.7 kB)
        Collecting sphinx-rtd-theme (from indic-nlp-library)
          Downloading sphinx_rtd_theme-3.0.2-py2.py3-none-any.whl.metadata (4.4 kB)
        Collecting morfessor (from indic-nlp-library)
         Downloading Morfessor-2.0.6-py3-none-any.whl.metadata (628 bytes)
        Requirement already satisfied: pandas in /usr/local/lib/python3.11/dist-packages (from indic-nlp-library) (2.2.2)
        Requirement already satisfied: numpy in /usr/local/lib/python3.11/dist-packages (from indic-nlp-library) (2.0.2)
        Requirement already satisfied: python-dateutil>=2.8.2 in /usr/local/lib/python3.11/dist-packages (from pandas->indic-nlp-library) (2.9.0.post0)
        Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.11/dist-packages (from pandas->indic-nlp-library) (2025.2)
        Requirement already satisfied: tzdata>=2022.7 in /usr/local/lib/python3.11/dist-packages (from pandas->indic-nlp-library) (2025.2)
        Collecting sphinx>=5.1.0 (from sphinx-argparse->indic-nlp-library)
          Downloading sphinx-8.2.3-py3-none-any.whl.metadata (7.0 kB)
        Collecting docutils>=0.19 (from sphinx-argparse->indic-nlp-library)
          Downloading docutils-0.21.2-py3-none-any.whl.metadata (2.8 kB)
        Collecting sphinxcontrib-jquery<5,>=4 (from sphinx-rtd-theme->indic-nlp-library)
         Downloading sphinxcontrib_jquery-4.1-py2.py3-none-any.whl.metadata (2.6 kB)
        Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.11/dist-packages (from python-dateutil>=2.8.2->pandas->indic-nlp-library) (1.17.0)
        Collecting sphinxcontrib-applehelp>=1.0.7 (from sphinx>=5.1.0->sphinx-argparse->indic-nlp-library)
          Downloading sphinxcontrib_applehelp-2.0.0-py3-none-any.whl.metadata (2.3 kB)
        Collecting sphinxcontrib-devhelp>=1.0.6 (from sphinx>=5.1.0->sphinx-argparse->indic-nlp-library)
          Downloading sphinxcontrib_devhelp-2.0.0-py3-none-any.whl.metadata (2.3 kB)
        Collecting sphinxcontrib-htmlhelp>=2.0.6 (from sphinx>=5.1.0->sphinx-argparse->indic-nlp-library)
          Downloading sphinxcontrib_htmlhelp-2.1.0-py3-none-any.whl.metadata (2.3 kB)
        Collecting sphinxcontrib-jsmath>=1.0.1 (from sphinx>=5.1.0->sphinx-argparse->indic-nlp-library)
          Downloading sphinxcontrib_jsmath-1.0.1-py2.py3-none-any.whl.metadata (1.4 kB)
        Collecting sphinxcontrib-qthelp>=1.0.6 (from sphinx>=5.1.0->sphinx-argparse->indic-nlp-library)
          Downloading sphinxcontrib_qthelp-2.0.0-py3-none-any.whl.metadata (2.3 kB)
        Collecting sphinxcontrib-serializinghtml>=1.1.9 (from sphinx>=5.1.0->sphinx-argparse->indic-nlp-library)
         Downloading sphinxcontrib_serializinghtml-2.0.0-py3-none-any.whl.metadata (2.4 kB)
        Requirement already satisfied: Jinja2>=3.1 in /usr/local/lib/python3.11/dist-packages (from sphinx>=5.1.0->sphinx-argparse->indic-nlp-library) (3.1.6)
        Requirement already satisfied: Pygments>=2.17 in /usr/local/lib/python3.11/dist-packages (from sphinx>=5.1.0->sphinx-argparse->indic-nlp-library) (2.19.1)
        Collecting snowballstemmer>=2.2 (from sphinx>=5.1.0->sphinx-argparse->indic-nlp-library)
          Downloading snowballstemmer-2.2.0-py2.py3-none-any.whl.metadata (6.5 kB)
        Collecting babel>=2.13 (from sphinx>=5.1.0->sphinx-argparse->indic-nlp-library)
          Downloading babel-2.17.0-py3-none-any.whl.metadata (2.0 kB)
        Collecting alabaster>=0.7.14 (from sphinx>=5.1.0->sphinx-argparse->indic-nlp-library)
          Downloading alabaster-1.0.0-py3-none-any.whl.metadata (2.8 kB)
        Collecting imagesize>=1.3 (from sphinx>=5.1.0->sphinx-argparse->indic-nlp-library)
          Downloading imagesize-1.4.1-py2.py3-none-any.whl.metadata (1.5 kB)
        Requirement already satisfied: requests>=2.30.0 in /usr/local/lib/python3.11/dist-packages (from sphinx>=5.1.0->sphinx-argparse->indic-nlp-library) (2.32.3)
        Collecting roman-numerals-py>=1.0.0 (from sphinx>=5.1.0->sphinx-argparse->indic-nlp-library)
          Downloading roman_numerals_py-3.1.0-py3-none-any.whl.metadata (3.6 kB)
   # Import All Required Libraries
    import gdown
    from google.colab import files
   from PIL import Image
    import numpy as np
    import cv2
    import torch
    import pytesseract
    import re
    import pandas as pd
    import matplotlib.pyplot as plt
    from deep_translator import GoogleTranslator
    from indicnlp.tokenize.indic_tokenize import trivial_tokenize
   from langdetect import detect
    !huggingface-cli clear-cache
    from huggingface_hub import login
   Step 1: Upload Image
   # Option for the user to try dynamic upload quickly
   user_choice = input("Would you like to skip static image and upload a dynamic image? (y/n) [default: n]: ").strip().lower()
    if user_choice == 'y':
       print("Proceeding with dynamic image upload...")
```

```
# Dynamic Image Uploads
uploaded = files.upload() # Upload image from user
filename = list(uploaded.keys())[0] # Get the name of the uploaded file
# Open image with PIL
img_pil = Image.open(filename)
img_pil.show()
# Static Image Uploads
file_id = '1z526YFcKb2g8HftFLPhH9Gg25I-jNtHh' # Extract the file ID from the shared link
url = f'https://drive.google.com/uc?export=download&id={file_id}'
   # Try downloading the static image using gdown
   gdown.download(url, 'marathi.gif', quiet=False)
   # Open the image with PIL
   img_pil = Image.open('marathi.gif')
    img_pil.show()
 except Exception as e:
   # If static image download fails, handle with dynamic image upload
   print(f"Static image download failed with error: {e}")
   print("Proceeding with dynamic image upload...")
   # Dynamic Image Uploads
   uploaded = files.upload() # Upload image from user
   filename = list(uploaded.keys())[0] # Get the name of the uploaded file
   # Open image with PIL
    img_pil = Image.open(filename)
    img_pil.show()
```

Would you like to skip static image and upload a dynamic image? (y/n) [default: n]: n Downloading... From: https://drive.google.com/uc?export=download&id=1z526YFcKb2g8HftFLPhH9Gg25I-jNtHh To: /content/marathi.gif 7.23k/7.23k [00:00<00:00, 14.4MB/s]

Step 2: Preprocess Image

```
# Convert PIL to OpenCV format
 img_cv = cv2.cvtColor(np.array(img_pil), cv2.COLOR_RGB2BGR)
# Preprocess image
gray = cv2.cvtColor(img_cv, cv2.COLOR_BGR2GRAY)
blurred = cv2.GaussianBlur(gray, (3, 3), 0)
 _, thresh = cv2.threshold(blurred, 0, 255, cv2.THRESH_BINARY + cv2.THRESH_OTSU)
#Show the preprocessed image
plt.imshow(thresh, cmap='gray')
plt.title("Preprocessed Image")
plt.axis('off')
plt.show()
```

हे प्रांत एकत्र जोडून मोगल आणि ब्रिपीशांना आव्हान देणा या स्वराज्याची स्थापना केली तेव्हा ख या अर्थाने महाराष्ट्राचा जन्म झाला. १६७४ साली शिवाजी महाराजांच्या राज्याभिषेकाने सुवर्णयुगाची नांदी झाली. त्यांच्या राजवटीच्या प्रत्येक दिवसाने शिवनेरी होनाला (राज्याभिषेकाच्या वेळी पाहलेली सोन्याची नाणीप झळाळी आणली. मराठयांनी स्वराजाच्या उज्वल परंपरेचे रक्षण केले आणि दोन तृतियांश भारत आपल्या ताब्यात आणला. नंतर पेशव्यांच्या काळात स्वराज्याला उतरती कळा लागली आणि शेवफी त्याचा अस्त झाला.

Preprocessed Image

१७ व्या शतकात जेव्हा छत्रपती शिवाजी महाराजांनी कोकण, मावळ आणि देश

→ Step 3: Transformer based OCR Extraction

✓ Trocr

 $\overline{\Rightarrow}$

from transformers import TrOCRProcessor, VisionEncoderDecoderModel

Load the pre-trained TrOCR model and processor processor = TrOCRProcessor.from_pretrained("microsoft/trocr-base-handwritten") model = VisionEncoderDecoderModel.from pretrained("microsoft/trocr-base-handwritten") model.eval() # inference mode only

Optional: use GPU if available device = torch.device("cuda" if torch.cuda.is_available() else "cpu")

https://colab.research.google.com/drive/1YgqR6DplHggogXh2Wv46oclJAHjUVnSG#scrollTo=Xq1VtB151UAN&printMode=true

```
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                                                                                                                                                                                                   NLP_TA2_Project.ipynb - Colab
    model.to(device)
   # Step 1: Load or convert the image
   # If using OpenCV image (NumPy array), convert to PIL
   if isinstance(img_pil, np.ndarray):
       img_pil_ = Image.fromarray(img_pil)
   # Step 2: Preprocess (resize and RGB)
    img_pil_ = img_pil.convert("RGB")
   img_pil_ = img_pil_.resize((384, 384))
   # Step 3: Feature extraction
   pixel_values = processor(images=img_pil_, return_tensors="pt").pixel_values.to(device)
   # Step 4: Generate text from image
   with torch.no_grad():
       generated_ids = model.generate(pixel_values)
       generated_text = processor.batch_decode(generated_ids, skip_special_tokens=True)[0]
    print("\n > OCR Text with TrOCR:\n", generated_text)
    /usr/local/lib/python3.11/dist-packages/huggingface_hub/utils/_auth.py:94: UserWarning:
        The secret `HF_TOKEN` does not exist in your Colab secrets.
        To authenticate with the Hugging Face Hub, create a token in your settings tab (<a href="https://huggingface.co/settings/tokens">https://huggingface.co/settings/tokens</a>), set it as secret in your Google Colab and restart your session.
        You will be able to reuse this secret in all of your notebooks.
        Please note that authentication is recommended but still optional to access public models or datasets.
          warnings.warn(
        preprocessor_config.json: 100%
                                                                        224/224 [00:00<00:00, 29.2kB/s]
        Using a slow image processor as `use_fast` is unset and a slow processor was saved with this model. `use_fast=True` will be the default behavior in v4.52, even if the model was saved with a slow processor was saved with `use_fast=False`.
                                                                      1.12k/1.12k [00:00<00:00, 157kB/s]
        tokenizer_config.json: 100%
                                                              899k/899k [00:00<00:00, 4.60MB/s]
         vocab.json: 100%
         merges.txt: 100%
                                                              456k/456k [00:00<00:00, 31.8MB/s]
         special_tokens_map.json: 100%
                                                                         772/772 [00:00<00:00, 113kB/s]
                                                              4.17k/4.17k [00:00<00:00, 636kB/s]
         config.json: 100%
         model.safetensors: 100%
                                                                   1.33G/1.33G [00:08<00:00, 151MB/s]
        Config of the encoder: <class 'transformers.models.vit.modeling_vit.ViTModel'> is overwritten by shared encoder config: ViTConfig {
          "attention_probs_dropout_prob": 0.0,
          "encoder_stride": 16,
          "hidden_act": "gelu",
          "hidden_dropout_prob": 0.0,
          "hidden_size": 768,
          "image_size": 384,
          "initializer_range": 0.02,
          "intermediate_size": 3072,
          "layer_norm_eps": 1e-12,
           "model_type": "vit",
           "num_attention_heads": 12,
          "num_channels": 3,
          "num_hidden_layers": 12,
           "patch_size": 16,
           "pooler_act": "tanh",
           "pooler_output_size": 768,
           "qkv_bias": false,
          "torch_dtype": "float32",
          "transformers_version": "4.51.1"
        Config of the decoder: <class 'transformers.models.trocr.modeling_trocr.TrOCRForCausalLM'> is overwritten by shared decoder config: TrOCRConfig {
          "activation_dropout": 0.0,
           "activation_function": "gelu",
           "add_cross_attention": true,
           "attention_dropout": 0.0,
           "bos_token_id": 0,
          "classifier_dropout": 0.0,
          "cross_attention_hidden_size": 768,
          "d_model": 1024,
           "decoder_attention_heads": 16,
           "decoder_ffn_dim": 4096,
           "decoder_layerdrop": 0.0,
          "decoder_layers": 12,
           "decoder_start_token_id": 2,
          "dropout": 0.1,
          "eos_token_id": 2,
          "init_std": 0.02,

    Abhi №64 Maha Phrase_maha BERTv2_Finetuning

          "layernorm_embedding": true,
          "may position ambaddings". E13
   # Use pytesseract for raw text
   ocr_text = pytesseract.image_to_string(thresh, lang="mar")
   # Run through 13cube-pune/marathi-bert-v2 for context-aware prediction
   from transformers import pipeline
   ocr_model = pipeline("text-classification", model="Abhi964/MahaPhrase_mahaBERTv2_Finetuning", tokenizer="Abhi964/MahaPhrase_mahaBERTv2_Finetuning")
   ocr_result = ocr_model(ocr_text)
   print("\n @ OCR Text:\n", ocr_text)
   print("\n \( \) MahaBERT Inference:\n", ocr_result)
        TOU SHOUTH PLOUBULY THATM CHIS MOUEL OH A NOWH-SCIEBM CASK TO DE BUTE TO USE IT TOL PLEUTCTIONS AND INTELENCE.
   genfigajtion:_ct@fi@j
                                                              728/728 [00190 ≮0900 0000 10000 $\dB0/$100, 26.4kB/s]
                                                                    950M/950M [00:06<00:00, 154MB/s]
         model.safetensors: 100%
OCR Text with Tround
        tdkeក/ze58config.json: 100%
                                                                      1.30k/1.30k [00:00<00:00, 185kB/s]
         vocab.txt: 100%
                                                            3.16M/3.16M [00:00<00:00, 31.5MB/s]
                                                                6.41M/6.41M [00:00<00:00, 7.56MB/s]
         tokenizer.json: 100%
                                                                         695/695 [00:00<00:00, 101kB/s]
        special_tokens_map.json: 100%
        Device set to use cpu
         0CR Text:
         १७ व्या शतकात जेव्हा छत्रपती शिवाजी महाराजांनी कोकण, मावळ आणि देश
        हे प्रांत एकज जोडून मोगल आणि ब्रिपीशांना आव्हान देणा या स्वराज्याची
        स्थापना केली तेव्हा ख या अर्थाने महाराष्यराचा जन्म झाला. ९६७४ साली
        शिवाजी महाराजांच्या राज्याभिषेकाने सुवर्णयुगाची नांदी झाली. त्यांच्या
        राजबटीच्या प्रत्येक दिवसाने शिवनेरी होनाला (राज्याभिषेकाच्या वेळी पाडलेली
        सोन्याची नाणीप झळाळी आणली. मराठयांनी स्वराजाच्या उग्जल परंपरेचे
        रक्षण केले आणि दोन तृतियांश भारत आपल्या ताब्यात आणला. नंतर
        पेशव्यांच्या काळात स्वराज्याला उत्तरती कळा लागली आणि शेबपी त्याचा
        अस्त झाला.
         MahaBERT Inference:
         [{'label': 'LABEL_0', 'score': 0.969126284122467}]
  Step 4: Normalize Text
   # Normalize the OCR output
   normalized_text = re.sub(r'[^\u0900-\u097F\s]', '', ocr_text)
    normalized_text = re.sub(r'\s+', ' ', normalized_text).strip()
    print("\nNormalized Marathi Text:\n", normalized_text)
        Normalized Marathi Text:
         १७ व्या शतकात जेव्हा छत्रपती शिवाजी महाराजांनी कोकण मावळ आणि देश हे प्रांत एकज जोडून मोगल आणि ब्रिपीशांना आव्हान देणा या स्वराज्यांच्या काळात स्व
  Step 5: Tokenization
```

Tokenization # Simple whitespace tokenizer tokens_simple = normalized_text.split() print("\nTokenized Words (Simple Split):\n", tokens_simple)

print("\nTokenized Words (Indic NLP):\n", tokens_advanced) tokens = tokens_advanced Tokenized Words (Simple Split): ['१७', 'व्या', 'शतकात', 'जेव्हा', 'छत्रपती', 'शिवाजी', 'महाराजांनी', 'साली', 'प्रोत', 'एकज', 'प्रोत', 'प्रात', Tokenized Words (Indic NLP):

['१७', 'व्या', 'शतकात', 'जेव्हा', 'छत्रपती', 'शिवाजी', 'महाराजांनी', 'महाराजांनी', 'पां, 'प्रतेक', 'देश', 'हे', 'प्रांत', 'एकज', 'जोडून', 'मोगल', 'पांज्याभिषेकाने', 'सुवर्णयुगाची', 'नांदी', 'महाराजांच्या', 'राज्याभिषेकाने', 'स्वराज्याची', 'स्वराज्याची', 'पांच्या', 'प्रतेव्हा', 'प्रांच्या', 'प्रांच्या', 'प्रतेव्हा', 'प्रांच्या', 'प्रांच्या', 'प्रतेव्हा', 'प्रांच्या', Step 6: Name Entity Recogination

NER Marathi

Advanced Indic NLP tokenizer

tokens_advanced = trivial_tokenize(normalized_text, lang='mar')

Use a pipeline as a high-level helper from transformers import pipeline pipe = pipeline("token-classification", model="13cube-pune/marathi-mixed-ner-iob") ner_results = pipe(normalized_text) print("\nNER Results:\n", ner_results) config.json: 100% 1.23k/1.23k [00:00<00:00, 178kB/s]

pytorch_model.bin: 100% 948M/948M [00:06<00:00, 156MB/s] model.safetensors: 100% 948M/948M [00:09<00:00, 144MB/s] 535/535 [00:00<00:00, 79.3kB/s] tokenizer_config.json: 100% 3.16M/3.16M [00:00<00:00, 38.0MB/s] vocab.txt: 100% 6.41M/6.41M [00:00<00:00, 34.6MB/s] tokenizer.json: 100% special_tokens_map.json: 100% 125/125 [00:00<00:00, 15.3kB/s] Device set to use cpu

NER Results: [{'entity': 'B-NEM', 'score': np.float32(0.9490036), 'index': 1, 'word': '१७', 'start': 0, 'end': 2}, {'entity': 'B-NED', 'score': np.float32(0.99822646), 'index': 5, 'word': '१० ', 'start': 29, 'end': 28, {'entity': 'B-NED', 'score': np.float32(0.99822646), 'index': 7, 'word': 'महारा

Step 7: Language Detection and Translation

Language Detection

Detect language detected_language_code = detect(normalized_text) # Map language code to full name (optional, for readability) language_map = { 'mr': 'Marathi', 'en': 'English', # Add more mappings as needed detected_language = language_map.get(detected_language_code, detected_language_code)

Print result print(f"Detected language: {detected_language}")

→ Detected language: Marathi

Translate individual tokens

Translation

Convert Marathi numbers to English numbers in text and tokens devanagari_to_english_digits = { '°': '0', '१': '1', '२': '2', '३': '3', '४': '4', '५': '5', '६': '6', '७': '7', '८': '8', '९': '9'

def convert_devanagari_numbers(text): return re.sub(r'[\u0966-\u096F]+', lambda m: ''.join(devanagari_to_english_digits.get(ch, ch) for ch in m.group()), text)

Translate full text normalized_with_english_digits = convert_devanagari_numbers(normalized_text) translation_text = GoogleTranslator(source='mr', target='en').translate(normalized_with_english_digits) print("\nEnglish Translated Text:\n", translation_text)

English Translated Text: In the 17th century, when Chhatrapati Shivaji Maharaj established the Konkan Maval and the country to challenge the Mughals and the British, Maharashtra was born in 9674 Shivaji Maharaj's coronation of the golden age. Protecting the tradition and two -thirds of India were taken into custody, after the Peshwa's time, Swarajya was responding and Sheeppi was dissolved.

translated_tokens = [] for token in tokens: if re.fullmatch(r'[\u0966-\u096F]+', token):

https://colab.research.google.com/drive/1YgqR6DplHggogXh2Wv46oclJAHjUVnSG#scrollTo=Xq1VtB151UAN&printMode=true

Step 8: Final Output

Display the image plt.imshow(img_pil)

plt.axis('off') plt.title("Uploaded Image") plt.show() # Final Output print("\nFinal Output Summary:\n") print("Original Marathi OCR Text:\n", normalized_text) # OCR Marathi Text print("\nMarathi to English Translation:\n", translation_text) # English Translated Text print("\nTokens in Marathi:\n", tokens) # Marathi Tokens print("\nTranslated Tokens in English:\n", translated_tokens) # English Tokens

Uploaded Image

१७ व्या शतकात जेव्हा छत्रपती शिवाजी महाराजांनी कोकण, मावळ आणि देश हे प्रांत एकत्र जोडून मोगल आणि ब्रिफीशांना आव्हान देणा-या स्वराज्याची स्थापना केली तेव्हा ख-या अर्थाने महाराष्ट्राचा जन्म झाला. १६७४ साली शिवाजी महाराजांच्या राज्याभिषेकाने सुवर्णयुगाची नांदी झाली. त्यांच्या राजवटीच्या प्रत्येक दिवसाने शिवनेरी होनाला (राज्याभिषेकाच्या वेळी पाडलेली सोन्याची नाणीप झळाळी आणली. मराठयांनी स्वराजाच्या उज्वल परंपरेचे रक्षण केले आणि दोन तृतियांश भारत आपल्या ताब्यात आणला. नंतर पेशव्यांच्या काळात स्वराज्याला उतरती कळा लागली आणि शेवफी त्याचा अस्त झाला.

Final Output Summary:

Original Marathi OCR Text: १७ व्या शतकात जेव्हा छत्रपती शिवाजी महाराजांनी कोकण मावळ आणि देश हे प्रांत एकज जोडून मोगल आणि ब्रिपीशांना आव्हान देणा या स्वराज्यांची स्थापना केली तेव्हा ख या अर्थाने महाराष्यराचा जन्म झाला ९६७४ साली शिवाजी महाराजांच्या राजबटीच्या प्रत्येक दिवसाने शिवनेरी होनाला राज्याभिषेकाच्या वेळी पाडलेली सोन्याची नाणीप झळाळी आणली मराठयांनी स्वराजाच्या उग्जल परंपरेचे रक्षण केले आणि दोन तृतियांश भारत आपल्या ताब्यात आणला नंतर पेशव्यांच्या काळात स्व

Marathi to English Translation: In the 17th century, when Chhatrapati Shivaji Maharaj established the Konkan Maval and the country to challenge the Mughals and the British, Maharashtra was born in 9674 Shivaji Maharaj's coronation of the golden age. Protecting the tradition and two -thirds of India were taken into custody, after the Peshwa's time, Swarajya was responding and Sheeppi was dissolved.

['१७', 'व्या', 'शतकात', 'जेव्हा', 'छत्रपती', 'शिवाजी', 'महाराजांनी', 'कोकण', 'मावळ', 'आणि', 'देश', 'हे', 'प्रांत', 'एकज', 'जेव्हा', 'शेवाजी', 'महाराजांच्या', 'राज्याभिषेकाने', 'स्वराज्याची', 'स्वराज्याची', 'स्वराज्याची', 'स्वराज्याची', 'साली', 'प्रांच्या', 'प्रांच् Translated Tokens in English:

['17', 'Th', 'Centuries', 'When', 'Chhatrapati', 'Shivaji', 'By the Maharaja', 'Konkan', 'Maval', 'And', 'Country', 'This', 'Province', 'Unity', 'By connecting', 'Maharashtra', 'Birth', 'Became', '9674', 'Sill', 'Shivaji', 'To the Maharaja', 'Coronat

Show tokens in tabular format df = pd.DataFrame({'Marathi': tokens, 'English': translated_tokens})

English Marathi 17 व्या 2 शतकात Centuries छत्रपती Chhatrapati

And शेबपी Shabby त्याचा Weed झाला Became 77 rows × 2 columns

Step 9: Sentiment Analysis

from transformers import pipeline

Sentiment Analysis

Try Marathi sentiment analysis model (if available) sentiment_pipeline = pipeline("sentiment-analysis", model="l3cube-pune/MarathiSentiment") sentiment_result = sentiment_pipeline(normalized_text) print("\nMarathi Sentiment Analysis Result:\n", sentiment_result)

print("\nMarathi sentiment model failed or not available. Falling back to English sentiment analysis.")

try: # Fallback: Use English-translated text and English sentiment model en_sentiment_pipeline = pipeline("sentiment-analysis") sentiment_result = en_sentiment_pipeline(translation_text) print("\nEnglish Sentiment Analysis Result:\n", sentiment_result)

except Exception as e: print("\nSentiment analysis failed due to:", str(e))

config.json: 100% 981/981 [00:00<00:00, 127kB/s] 134M/134M [00:00<00:00, 158MB/s] model.safetensors: 100% 442/442 [00:00<00:00, 43.0kB/s] tokenizer_config.json: 100% 5.65M/5.65M [00:00<00:00, 132MB/s] spiece.model: 100% 244/244 [00:00<00:00, 34.7kB/s] special_tokens_map.json: 100% Device set to use cpu

Marathi Sentiment Analysis Result: [{'label': 'Neutral', 'score': 0.983304500579834}]

Step 10: Summarize the Marathi and English text

Summarization using transformers pipeline summarizer_mar = pipeline("summarization", model="Existance/mT5_multilingual_XLSum-marathi-summarization") summarizer_en = pipeline("summarization", model="Falconsai/text_summarization")

try: marathi_summary = summarizer_mar(normalized_text, max_length=130, min_length=30, do_sample=False) print("\nMarathi Text Summary:\n", marathi_summary[0]["summary_text"]) except Exception as e:

try: english_summary = summarizer_en(translation_text, max_length=130, min_length=30, do_sample=False)

print(f"\nError summarizing Marathi text: {e}")

print("\nEnglish Text Summary:\n", english_summary[0]["summary_text"]) except Exception as e:

print(f"\nError summarizing English text: {e}") config.json: 100%

special_tokens_map.json: 100%

2.33G/2.33G [00:14<00:00, 159MB/s] pytorch_model.bin: 100% 2.33G/2.33G [01:12<00:00, 29.4MB/s] model.safetensors: 100% generation_config.json: 100% 234/234 [00:00<00:00, 29.1kB/s] 285/285 [00:00<00:00, 36.3kB/s] tokenizer_config.json: 100% spiece.model: 100% 4.31M/4.31M [00:00<00:00, 125MB/s] tokenizer.json: 100% 16.3M/16.3M [00:00<00:00, 156MB/s] 74.0/74.0 [00:00<00:00, 9.77kB/s] special_tokens_map.json: 100% Device set to use cpu 1.49k/1.49k [00:00<00:00, 177kB/s] config.json: 100% model.safetensors: 100% 242M/242M [00:01<00:00, 154MB/s] 112/112 [00:00<00:00, 14.6kB/s] generation_config.json: 100% 2.32k/2.32k [00:00<00:00, 256kB/s] tokenizer_config.json: 100% 792k/792k [00:00<00:00, 59.9MB/s] spiece.model: 100% 2.42M/2.42M [00:00<00:00, 37.6MB/s] tokenizer.json: 100%

Device set to use cpu Your max_length is set to 130, but your input_length is only 105. Since this is a summarization task, where outputs shorter than the input are typically wanted, you might consider decreasing max_length manually, e.g. summarizer('...', max_length=52)

मराठयांनी स्वराजाच्या उर्ग्जल परंपरेचे रक्षण केले आणि दोन तृतियांश भारत आपल्या ताब्यात आणला नंतर पेशव्यांच्या काळात स्वराज्याला उत्तरती कळा लागली आता शेबपी त्याचा अस्त झाला

2.20k/2.20k [00:00<00:00, 291kB/s]

908/908 [00:00<00:00, 110kB/s]