



GLOBAL ACADEMY OF TECHNOLOGY
DEPARTMENT OF COMPUTER SCIENCE AND ENGINEERING
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|-----------------------|-------------------------------------|---------------------|-------------------|
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| Domain | DEEP LEARNING | Group No: | 11 |
| Project Title | “ LANDSCAPE RECOGNITION ” | | |
| Under taken at | GLOBAL ACADEMY OF TECHNOLOGY | | |
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Agenda

- Problem Description
- Literature Survey
- Gaps in the literature survey
- Proposed Solution
- Requirements Specifications
- Project Work plan
- Bibliography

Problem Description

- In the field of Artificial Intelligence (AI) and Computer Vision recognition of objects has become very common, feasible and realistic.
- Looking ahead, there will come a time where instance-specific recognition will become a trend and be an everyday problem.

- Identifying the instances with the help of present day techniques is still primitive as there is a huge spectrum of features.



Fig: 1 - Eiffel Tower and Tokyo Tower



Fig: 2 - India Gate and Arc De Triomphe

Literature Survey

| Title of the paper and year | Methodology | Advantages | Disadvantages |
|---|---|--|---|
| Rich feature hierarchies for accurate object detection and semantic segmentation - 2014 | our system <ul style="list-style-type: none"> • takes an input image • extracts around 2000 bottom-up region proposals • computes features for each proposal using a large convolutional neural network (CNN), and then • classifies each region using class-specific linear SVMs | <ul style="list-style-type: none"> • provides higher accuracy than CNNs (R-CNN achieves a mean average precision (MAP) of 53.7% on PASCAL VOC 201 for comparison, reports 35.1% MAP) | <ul style="list-style-type: none"> • training is multi-stage pipeline • training is expensive in time and space • object detection is slow |
| Fast R-CNN - 2015 | <ul style="list-style-type: none"> • the image is processed with several convolutional and max pooling • then region of interest has been extracted from feature map • each feature vector is fed into a fully connected layers that finally branches into two outputs | <ul style="list-style-type: none"> • training is single-stage, using a multi-task loss • training can update all network layers • no disk storage is required for feature caching | <ul style="list-style-type: none"> • most of the time taken by Fast R-CNN during detection is a selective search region proposal generation algorithm. Hence, it is the bottleneck of this architecture. |
| Large-Scale Image Retrieval with Attentive Deep Local Features - 2017 | <ul style="list-style-type: none"> • extract dense features from an image by applying a fully convolutional network • using RANSAC and employ the number of inliers as the score for retrieved images | <ul style="list-style-type: none"> • DELF clearly outperforms all other techniques significantly • DELF has higher recall • Attention helps more than fine-tuning | <ul style="list-style-type: none"> • pipeline requires less than 8GB memory to index 1 billion descriptors • challenges in query image with no correct match |

Gaps in the literature survey

Even though all the papers try to perform image detection there are still few challenges which are not completely eradicated by those methods

- Partial occlusion
- Multiple landmarks
- Queries with no match
- Local features lack semantic information
- Patch-level annotations are expensive
- Existing datasets are small/medium

Proposed Solution

Our solution approach can be decomposed into 2 main blocks:

- i. Localized feature extraction
 - ii. Querying and Retrieval.
- The Localized feature extraction block is formed with a CNN architecture with feature extracting layers trained with a classification loss.
 - Next, we will aggregate all the matches per database image. Finally, we perform geometric verification using RANSAC and employ the number of inliers as the score for retrieved images.

Requirements Specification

- **Hardware Requirements**

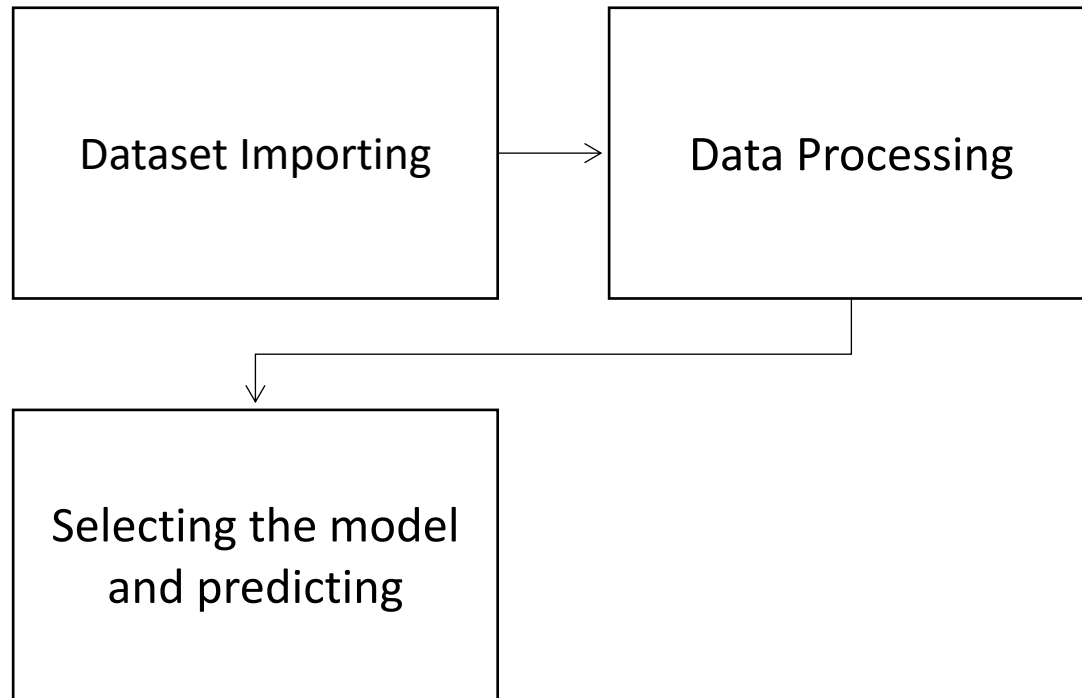
- Intel i5 8th gen
- 8gb ram
- 2gb graphics card
- 500gb HDD

- **Software Requirements**

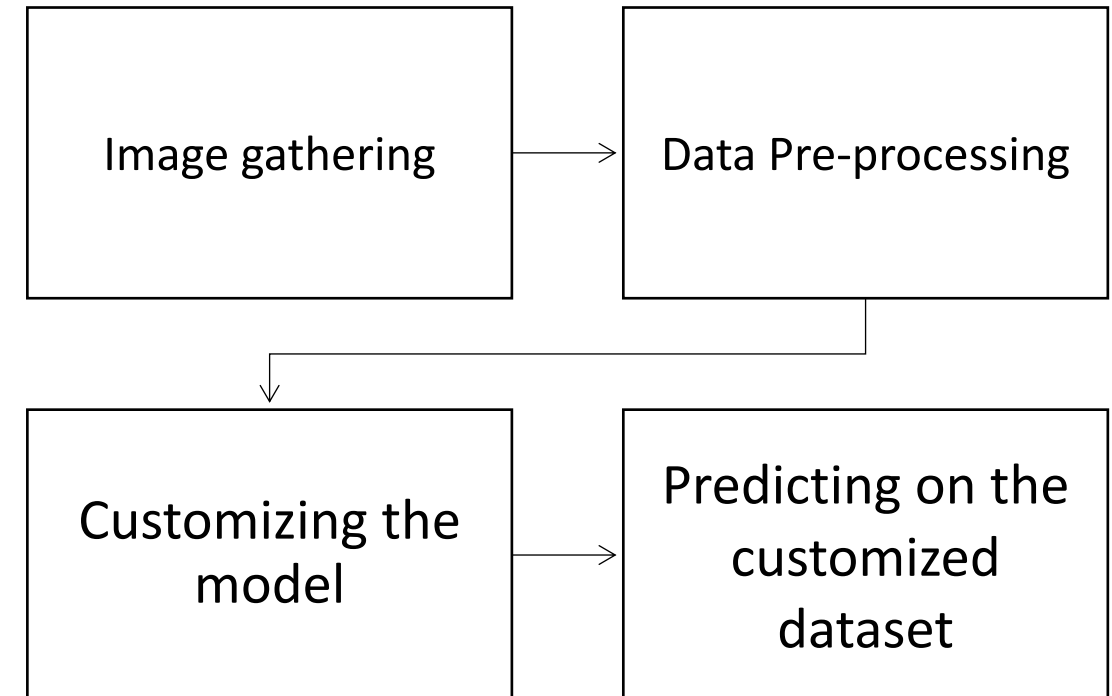
- *Python 3.8*
- *TensorFlow 2.x*
- Anaconda 3

Project Work plan

Project Phase 1



Project Phase 2



Conclusion of the presentation

- This solution can prove that instance level recognition is achievable and can be put into practical use.
- This Project can pave way for undiscovered roads in the field of Computer Vision and Object Recognition.

Bibliography

- [1] Hyeonwoo Noh, Andre Araujo, Jack Sim, Tobias Weyand, Bohyung Han. “Large-Scale Image Retrieval With Attentive Deep Local Features”. IEEE, 2017, pp. 3456-3465
- [2] Ross Girshick, Jeff Donahue, Trevor Darrell, Jitendra Malik. “Rich feature hierarchies for accurate object detection and semantic segmentation tech report (v5)”. IEEE, 2014
- [3] Ross Girshick. “Fast R-CNN”. IEEE, 2015, pp. 2380-7504

Q & A

Thank You