**caMicroscope**

Proposal - Google Summer of Code 2020

**Project Proposal**

1. **Project Title**

Cancer Region of Interest Extraction and Machine Learning

1. **Abstract**

From what I have gathered from my research over the internet, caMicroscope is an open-source digital pathology data management, visualization and analysis platform. As of now, It supports single-stage networks showing results directly and is basically is a tool to view, annotate, and analyze biomedical images.

After going through the public website of [caMicroscope](https://wolf.cci.emory.edu//camic_org/apps/landing/landing.html), I got the gist of the work that they do. Generally for such projects, Histopathological images are used from datasets like The Breast Cancer Histopathological Image Classification (BreakHis) and classes are created like “Benign” and “Malignant” for the detection of cancer.

What this project seeks is that once a region of interest is extracted from the image it should be downloaded and fed to other models for tasks like synthetic data generation. The aim is to support a wider range of models and flexible use of their outputs.

1. **Student Details**

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1. **Potential Mentors**

Insiyah Hajoori and Ryan Birmingham

1. **Personal Background**

I am a third-year Bachelors of Technology Undergraduate student of Computer Science Engineering with an interest in Neural Networks.

**Projects / Freelancing / Internships**

|  |  |  |  |
| --- | --- | --- | --- |
| **Duration**  **(From –**  **till)** | **Organization** | **Title** | **Concepts/Technology used** |
| January 2020 –  Present | Manipal  University Jaipur | Occlusion detection  and gait silhouette  reconstruction from  degraded scenes | Deep Learning Pipeline,  Convolutional Neural Networks(Alexnet, VGG 16), Yolov3, Python, Tensorflow, Tensorflow.js, Flask, Transfer Learning |
| October 2019 -  Present | Barrow  (My own Startup) | Food Ordering and  Expense Tracker App | Flutter, Python |
| July 2019– Dec  2019 | Freelance | Course and  Assignments sharing  platform for the  faculty of universities | Used Reactjs for Frontend |
| May 2019 –  Jul 2019 | IITD-AIA Foundation  for Smart  Manufacturing  (IAFSM) | Industrial IoT 4.0 Research and  Development  Internship | • Designed, Developed and  Deployed an IoT  Dashboard for monitoring  industrial systems.  • The service was linked with  real-time Firebase database  • Hands-on experience in  OPC Unified Architecture  and Node-RED |

1. **Code Challenge/Work done till now**

We were given the challenge to make a project on interest extraction Using a machine learning toolkit of our choice. The project was to create a tool which identifies objects in the image, then returns positions in pixels corresponding to bounding boxes of the object in the image.

I started on this project on the 23rd of February. Here are my error/progress logs:

LOGS START

**23rd February 2020**

Started my research on various other CNN architectures and object detection models.

**29th February 2020**

Created a workflow for me. Divided my project into:

1. Architecture Design
   1. Create the layers of the network architecture
   2. Implement the forward pass of the network
   3. Objectness Confidence Thresholding and Non-maximum Suppression
   4. Design the input and the output pipelines
2. Data collection
   1. kaggle
3. Data preparation (even though I’ll use Kaggle)
   1. Verify the annotations are correct
   2. Ensure the EXIF orientation of the images is correct
   3. Resize images and update image annotations to match the newly sized images
   4. Colour corrections
   5. Format annotations to match the requirements of the model’s inputs
4. Model training and improvement
   1. Will use Google colab
   2. Load in the data via the Kaggle
   3. Determine the model configuration
   4. Initiate training
   5. Use the trained model for inference
5. Deploy using Flask and firebase
6. Give a front end
7. Detect!

**2nd March 2020**

Used the official cfg file from [the author](https://raw.githubusercontent.com/pjreddie/darknet/master/cfg/yolov3.cfg) for the network.

Created darknet.py. Parsing through cfg.

5 types of layers: Convolutional, Shortcut, Upsample, Route, YOLO.

Extended the nn.Module nn.Sequential and torch.nn.parameter classes.

#### Implemented the forward pass of the network

#### Testing the forward pass: failed

Cant store the image.

**4th March 2020**

Testing forward pass:

model = Darknet("cfg/yolov3.cfg")

inp = get\_test\_input()

pred = model(inp, torch.cuda.is\_available())

print (pred)

( 0 ,.,.) =

16.0962 17.0541 91.5104 ... 0.4336 0.4692 0.5279

15.1363 15.2568 166.0840 ... 0.5561 0.5414 0.5318

14.4763 18.5405 409.4371 ... 0.5908 0.5353 0.4979

⋱ ...

411.2625 412.0660 9.0127 ... 0.5054 0.4662 0.5043

412.1762 412.4936 16.0449 ... 0.4815 0.4979 0.4582

412.1629 411.4338 34.9027 ... 0.4306 0.5462 0.4138

[torch.FloatTensor of size 1x10647x85]

Downloaded weights and loaded them with load\_weights

model = Darknet("cfg/yolov3.cfg")

model.load\_weights("yolov3.weights")

As of now: built the model and weights loaded

Can finally start detecting objects.

**10th March 2020**

Finally started with detection. Wrote app.py

**12th march 2020**

Debugged app.py

Deployed it using flask

**15th March 2020**

Finally got the image on the server. Used Pyrebase4

Do not use Pyrebase

LOGS END

**7. Project Schedule**

1. 28th April - 18th May: **Community Bonding**
   1. Community Bonding.
   2. Understanding the structure of caMicroscope.
   3. Getting familiar with all the relevant technologies/ concepts.
2. 19th May - 10th June: **Development Phase**
   1. Development starts
   2. Reviewing the whole architecture
   3. Making the necessary additions to the pre-existing model
3. 15th June - 19th June: **Phase 1 Submission**
   1. submitting Phase 1 evaluations
   2. Inline comments
   3. Comment on a selection
4. 20th June - 13th July - **Development Phase**
   1. Development resumes
   2. Reviewing the whole architecture
   3. Making the necessary additions to the pre-existing model
5. 13th July - 17th July - **Phase 2 Submission**
   1. submitting Phase 2 evaluations
   2. Inline comments
   3. Comment on a selection
6. 18th July - 31st July - **Allow admin to limit access to specific files**
   1. Up to this point, a user will have common permission (View/Edit/Comment) to access all files.
   2. Add feature to allow users to provide a different level of access to different files.
7. 1st Aug - 9th Aug -**Project Completion, testing, and documentation**
   1. Wrapping up the project
   2. Documenting
8. 10th Aug - 17th Aug - **Final Testing**
   1. Final evaluation

**8. Planned absence/vacation days and other commitments during the GSoC period**

I will be having my university summer vacations from May 16th Onwards till July 24th 2020. No plans for any vacations. Working hours are shown below.

1. **Planned GSoC work hours**

My prefered working hours are:

|  |  |  |
| --- | --- | --- |
| **Indian Standard Time (IST)** | **Coordinated Universal Time(UTC)** | **Total Time** |
| 11:30am-13:30pm | 6:00am-8:00am | 2 hours |
| 15:00pm-19:00pm | 9:30am-13:30pm | 4 hours |
| 22:00pm-05:00am | 16:30pm-23:30pm | 7 hours |

1. **Skill Set**

I have done Courses on Neural Networks and Convolutional Neural Networks from Coursera.

Have worked on projects with Tensorflow and Tensorflow.js

Currently, I am working on GAIT analysis: Occlusion detection and gait silhouette reconstruction from degraded scenes using deep learning wherein I have prepared a customized layer architecture for our dataset provided by the Technical University of Munich and IIT Kharagpur (TUM-IITKPG) following is the gist of the work:

Stage 1: Background removal

In image processing there usually is preprocessing of the image where the first step is segmentation. This refers to where the background is separated from the foreground to isolate the foreground.

The background removal techniques:

1. Non-recursive technique

2. Recursive technique

Stage 2: Feature extraction

Once the background is subtracted from the image, each image sequence is converted into a temporary sequence of distance signals. A feature of an image is represented using a vector.

Two approaches for feature extraction

Approach 1: Model-based approach

* high-quality video sequences
* parameters used: height, the distance between the pelvis and feet and the distance between the feet.
* The silhouette is divided into some regions that are feature vectors that include averages of the centroid, the aspect ratio and orientation of the major axis of the ellipse

Approach 2: Holistic approach

* More focus on the shape of the silhouette or the motion of the whole body as compared to model-based approaches that focus on a particular part of the body.
* not dependent on the quality of the video frame.
* offer less computational requirements and complexities
* Not as robust as compared to model-based approaches
* contour of the silhouette is regarded as an important factor of the method.
* For high-quality silhouette, the outer contour of the silhouette is regarded as the important feature
* For low-quality silhouette, the binaries silhouettes are regarded as the important feature

Methodology

* Scrutinizing the data
* Automated identification of humans from their gait is a challenging research problem.
* We consider situations where the subject gets occluded due to the presence of multiple objects in the field of view of the camera, which is quite common in real-world surveillance scenarios.
* There are two types of occlusions: Static and Dynamic.
* Upon research, we came across a novel approach for detecting the presence of occlusion in a sequence of silhouette frames and their subsequent reconstruction.
* Dynamic programming- based maximum likelihood key pose detection algorithm simultaneously detects key pose class for each frame and also identifies the occluded frames.
* Clean and unclean gait cycles are segregated as the algorithm output. If all the subsequences of frames corresponding to a gait cycle are degraded by occlusion, then none of the existing methods can be used for recognition from this sequence.
* Need for the reconstruction of the degraded silhouette frames to construct clean gait cycles becomes pertinent in such situations.
* A novel method based on BGPDM which is able to reconstruct the missing silhouettes considerably well.
* The reconstructed silhouettes can then be used for recognition using any of the existing methods. We will test our algorithms on the data set (TUM-IITKGP) featuring occlusion by static objects as well as dynamic objects.