Weekly Schedule Schedule: Sunday, 2:00 PM

Our github repo: [alckasoc/Team-Chiken-wi22: The official ACM AI Team Chiken repository. (github.com)](github.com)

2/27/2022 Meeting Time: Sunday, 2:00 pm
Attendees: Nathan, Derrick, Min, David, Vincent

Summary of Meeting

- Progress checks on everyone
    - Everyone is still learning the ropes of how everything works (which is fine)!
- Pitch new project idea (sentiment analysis + OCR)
- Covered what we will be doing going forward

Action Items

1. Learn easyOCR (figure out how it works) -> google stuff, look at docs, watch tutorials, experiment with the code, etc
2. Go through a basic sentiment analysis pipeline -> you can find this on tensorflow docs (or generally anywhere; this is a pretty popular first NLP project)

Timeline

1. Continue (and eventually finalize) technical process
    a. First point, understand OCR (generally speaking) and understand sentiment analysis with deep NLP
    b. After you get your fundamentals down, apply your knowledge and learned practical skills to a custom dataset (y'all choose)
    c. Attach that OCR component into your customized sentiment analysis pipeline
    d. Maybe tune things, experiment, play around with the code!
2. Expand our horizons and turn this into an app (if time allows)
    a. Not sure if we have time, but if we do and we have the relevant expertise, then we can definitely beautify this project
3. ACM Project Showcase
    a. ACM will be having a project showcase (which is ages from now, but still nice to include here)
    b. Slideshow, visuals, demos, and thorough explanations of your entire project from when you started to the finished product!
    c. Talk about your difficulties, your many approaches, what you've done, what y'all have discussed, how you fixed issues, what you might do going forward, etc.

Timeline - Technical Process
(This is my sketch of the project)

1. Flush out the problem statement
    a. Are we using deep learning, if so what aspect of deep learning?

      b.   Is this problem feasible? Will it require rare data?

      c.   Where do we find the data?

      d.   Is this problem suited for AI?

2. Find reliable dataset(s)
   a. Resources: Kaggle, UCI, other ones available online
   b. Is this dataset suited for our task?
   c. What is it missing? Pros and cons?

3. Wrangle the data
   a. Are there missing values? Are there missing features?
   b. Are there some underlying problems in the dataset?

4. Preprocess the data
   a. Transform the data into a format your model will like!

5. Modeling
   a. Custom model or use off-the-shelf/ready-to-use models
   b. What model should we be using? Why did you pick that model? Pros and cons?
   c. Just from learning about how the model works, how do we improve this model? What are its weaknesses?

6. Inference/Validation
   a. Inference and validation are different things!
   b. Validation is tied to the model training loop and inference is getting a few new samples of predictions from the model
   c. I put both of them here because you monitor your model's performance through validation and also sometimes inference
   d. Let's see some results and samples.
   e. Dive deeper into its weaknesses and perform some sort of error analysis!

7. Repeat
   a. There are many components to your ML pipeline (points 2-6 above) and also many other subcomponents (that may be excluded from the above points)
   b. Find a concrete way to diagnose the weaknesses of your *pipeline* (not model but the entire pipeline from ingesting the data to making predictions) and adjust it accordingly
      i. Maybe the model is underfitting or the hyperparameters are wack
      ii. Maybe the dataset is poor in quality
      iii. Maybe your preprocessing pipeline needs to be improved
      iv. etc