

# Project Flow

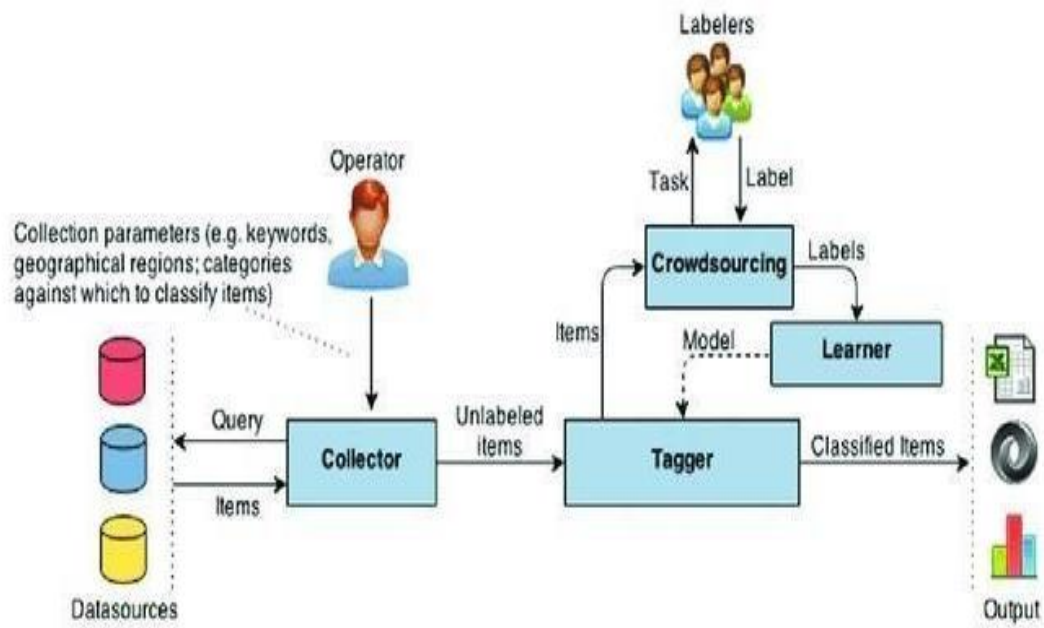
<b>Date</b>	<b>14<sup>th</sup> September 2022</b>
<b>Team ID</b>	<b>PNT2022TMID54341</b>
<b>Project Name</b>	<b>Natural Disasters Intensity Analysis and Classification Using Artificial Intelligence</b>

- The user interacts with the UI (User Interface) and give the image as input.
- Then the input image is then pass to our flask application,
- And finally with the help of the model which we build we will classify the result and showcase it on the UI.

To accomplish this, we have to complete all the activities and tasks listed below

- Data Collection.
- Collect the dataset or create the dataset
- Data Pre-processing.
- Import the Image Data Generator library
- Configure Image Data Generator class
- Apply Image Data Generator functionality to Trainset and Test set
- Model Building
- Import the model building Libraries
- Initializing the model
- Adding Input Layer
- Adding Hidden Layer
- Adding Output Layer
- Configure the Learning Process
- Training and testing the model
- Save the Model
- Application Building
- Create an HTML file
- Build Python Code

- Aerial imagery captured via unmanned aerial vehicles (UAVs) is playing an increasingly important role in disaster response.
- Unlike satellite imagery, aerial imagery can be captured and processed within hours rather than days.
- In addition, the spatial resolution of aerial imagery is an order of magnitude higher than the imagery produced by the most sophisticated commercial satellites today.
- Both the United States Federal Emergency Management Agency (FEMA) and the European Commission's Joint Research Center (JRC) have noted that aerial imagery will inevitably present a big data challenge.
- The purpose of this article is to get ahead of this future challenge by proposing a hybrid crowd sourcing and real-time machine learning solution to rapidly process large volumes of aerial data for disaster response in a time-sensitive manner.
- Crowdsourcing can be used to annotate features of interest in aerial images (such as damaged shelters and roads blocked by debris).
- These human-annotated features can then be used to train a supervised machine learning system to learn to recognize such features in new unseen images.
- The hybrid solution we present can be applied to both aerial and satellite imagery and has applications beyond disaster response such as wildlife protection, human rights, and archeological exploration.
- As a proof of concept, we recently piloted this solution using very high-resolution aerial photographs of a wildlife reserve in Namibia to support rangers with their wildlife conservation efforts
- The results suggest that the platform we have developed to combine crowdsourcing and machine learning to make sense of large volumes of aerial images can be used for disaster response.



Submitted By

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