

Resources

GitHub repo: https://github.com/CrowdTruth/CrowdTruth-core

Step-by-step getting started guide: https://git.io/fAAn7

Exercises Session 3: https://git.io/fAAc0

If you have working **Python** & **Jupyter Notebooks** installations:

pip install crowdtruth installs Python package

```
git clone git@github.com:CrowdTruth/CrowdTruth-core.git
cd CrowdTruth-core/
python setup.py install
```

If you don't have Python & Jupyter Notebooks, but you do have a Google account, use **Google Colab**:

- save the Colab notebooks to your personal Google Drive
- execute the notebooks straight from Drive

Step 1: Define the pre-processing configuration

```
import crowdtruth
from crowdtruth.configuration import DefaultConfig

class TestConfig(DefaultConfig):
   ...
```

The pre-processing configuration defines how to interpret the raw crowdsourcing input into an **annotation vector** that can be processed by the CrowdTruth metrics.

The configuration is defined as a **class**, inheriting from the Default configuration.

Configuration Class Attributes

```
import crowdtruth
from crowdtruth.configuration import DefaultConfig
class TestConfig(DefaultConfig):
  inputColumns = [...] #list of input columns from the .csv file with the input data
 outputColumns = [...] # list of output columns with the answers from the workers
 open ended task = ... # boolean var, whether or not task is open-ended
  annotation vector = [...] # list of possible annotations for closed tasks
 csv_file_separator = "," # column file separator
  annotation_separator = "," # output column annotation separator
 none token = "NONE" # name of annotation vector component for no answer picked
  remove empty rows = True # whether to remove empty judgments from the data
 # method where any additional processing of the raw judgments is done
  def processJudgments(self, judgments):
    return judgments
```

Amazon Mechanical Turk and Figure Eight (former Crowdflower) output files are automatically processed by the package.

Custom file types can be processed using the customPlatformColumns attribute:

```
class TestConfig(DefaultConfig):
  customPlatformColumns = [
   # column names in the custom .csv file,
   # must be declared in this specific order
    "{\$judgment_id}", # judgment ID column name
    "{$unit_id}", # unit ID column name
    "{$worker_id}", # worker ID column name
    "{$started_time}", # judgment start time
    "{$submitted time}" # judgment submit time
```

Step 2: Pre-process the raw data from the crowd to get the annotation vectors

Annotation vectors are now in data["judgments"]["output.X"], where X is the output column defined in the configuration.

Step 3: Calculate the CrowdTruth metrics

```
results = crowdtruth.run(data, config)
```

The CrowdTruth quality metrics are stored in:

- results["units"] input unit quality metrics
 - results["units"]["uqs"]: unit quality score
 - results["units"]["unit_annotation_score"]: ratio of workers that picked the annotation vs. all workers, weighted by worker quality
- results["workers"] worker quality metrics
 - results["workers"]["wwa"]: worker-worker agreement
 - results["workers"]["wsa"]: worker-unit agreement
 - results["workers"]["wqa"]: worker quality score
- results["annotations"] annotation quality metrics
 - results["annotations"]["aqs"]:annotatio

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- 1. Install CrowdTruth package & learn how to run
- 2. Explore notebooks implementing CrowdTruth metrics for different tasks
 - for local Jupyter server, download tutorial folder https://git.io/fAAC3
 - for Google Colab, follow links in https://git.io/fAAc0
 - ⇒ □ save .csv input as Google Sheets file + use Python 2 environment
 - if you don't want to code, analyze input & results .csv files (links in notebooks)
- 3. Compare an **open-ended** vs. **closed** config for the same task
- 4. Observe the effect of **dimensionality reduction** techniques over the quality metrics
- 5. Implement the annotation vector from Session 2