# Model Objective Validation

Avoiding risks associated with Design Objectives and Bias in Data



## **Objectives**

- How to determine the right set of metrics according to your business problem.
- How to build a baseline by using the defined metrics.
- How to identify weaknesses on the dataset. Evaluating fairness and representation on dataset.
- How to determine quality samples vs. target. i.e. miss-labeled samples.
- How to determine your Dev/Test strategies to ensure robustness of evaluation.
- Build baseline models.
- How to identify cases where you need to enrich dataset.

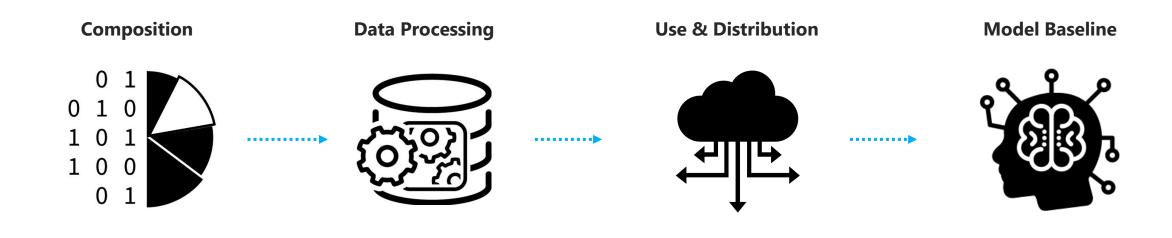


## Agenda

- Data Composition
- Data preprocessing or pre-formatting.
  (Before Feature Engineering)
- Data Use
- Model Baseline
- Questionnaire



## Data Ingestion | Composition

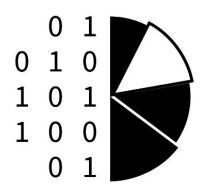


Before Feature Engineering



### Data Ingestion | Composition

#### **Composition**



- Does the dataset contain data that might be considered sensitive in any way?: (e.g., data that reveals racial or ethnic origins, sexual orientations, religious beliefs, political opinions or union memberships, or locations; financial or health data; biometric or genetic data)
- Does the dataset identify any subpopulations? If yes, please identify those variables: This question can help in the identification of possible biases or under-representation of certain subpopulations. Age, Sex, Gender and Race are examples.
- Is the dataset self-contained, or does it link to or otherwise rely on external resource? The answer here can be additional websites, tweets or other datasets.



### Data Ingestion | Collection Process

#### **Data Processing**

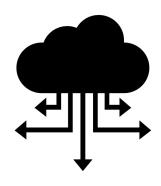


- Do you have the documentation preprocessing/cleaning/labeling of the data? If so, provide link to code: This question is based on the pre-processing of the data before storage. This is independend from the data preprocessing during Feature Engineering.
- Was the "raw" data saved in addition to the preprocessed/cleaned/labeled data? (e.g., to support unanticipated future uses)? This only applies to changes in data to be stored.
- Are there tasks for which the dataset should not be used?: Please specify whether this data cannot be used in other processes. This can be because several reasons: regulations, data not being adequate for other processes, etc...



## Data Ingestion | Storage

#### **Use & Distribution**



- Did you identify the set of metrics that will translate the business problem into Machine Learning?: Specify whether the data is currently being used by other process or model. This is important, as any change that you do to this dataset might affect other models.
- What is the structure of data?: What is the structure of the dataset? *Tabular?*, Non-Tabular? Graph\*. More information on data types here



### Data Ingestion Collection Process

#### **Model Baseline**



- Did you identify the set of metrics that will translate the business problem into Machine Learning?: This is very important; we need to have a way of measuring the improvement of the Machine Learning models.
- Do you have a business explanation why you chose those metrics?: It is important to have performance metrics that translated your business problems into an optimization problems, so that each improvement in those metrics can be translated to actual improvements in your processes.
- How are you evaluating the overall consistency labels (target) vs. input data? How are you identifying wrong labels or outputs?: it is important to identify possible outliers, you can do this by using an anomaly detection model.



## **Data Science Questionnaire for Data Ingestion**

Link:

