

# Machine Learning

# Sampling Techniques

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# Sampling

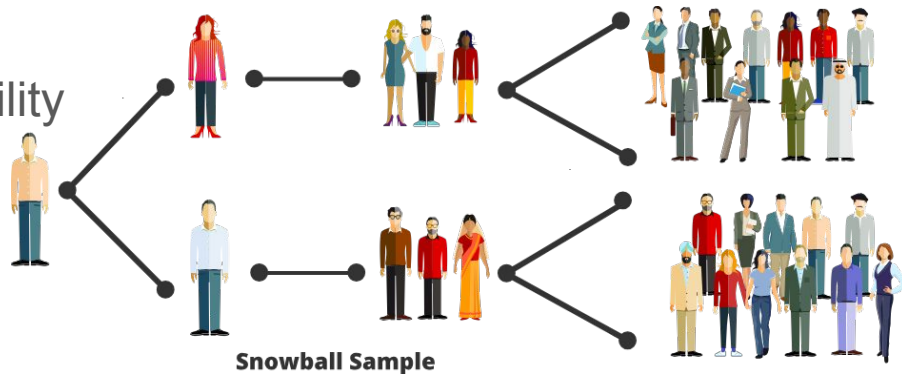
- Sampling techniques in machine learning are strategies used to select a **subset** of data from a larger dataset, typically for the purpose of **training or evaluation**.
- Different sampling techniques serve various **objectives**, such as balancing class distribution (oversampling / undersampling), reducing computational time, or improving model performance
- Sampling techniques may create a **bias** (such as examples selection in active Learning).
  - One has to carefully understand what kind of bias I am introducing and its effect

# Convenience Sampling

- Convenience sampling is a **non-probability** sampling technique where subjects are **selected** because of their convenient accessibility.
- This is the most common case for building early iterations of your dataset for POC
  - Easy and cheap to build, but clearly biased and can't generalize
  - For example, you collected data for in-cabin automotive in the summer using the workers in nearby factory who are mainly black males

# Snowball Sampling

- Non-probability sampling: Start from a few samples and from them identify more samples and so on
- For example, start from a few twitter accounts and use their connections to find more
  - Or Start from a page of links and crawl more pages
  - You always limited to the possible scope of early sample
  - Good for Social Network Analysis
  - Again biased
- Many datasets for follows non-probability sampling, such as crawling wiki or reviews (amazon/IMDM)



# Random Sampling

- **Simple** Random Sampling
  - Each instance has an equal chance of being selected.
  - Very common
- **Stratified** Random Sampling
  - Preserves the same class distribution as in the original dataset.
  - Typically in imbalance datasets
- **Weighted** sampling (with/out Replacement): assign weight for samples
  - `import` random
  - `items = ['apple', 'banana', 'cherry', 'date']`
  - `weights = [0.5, 0.2, 0.2, 0.1]`
  - `sampled_items = random.choices(items, weights, k=5)`

# Reservoir Sampling

- Imagine a stream of data coming to production (e.g. video frames from self-driving)
- We want to fine-tune our model on randomly selected  $K$  frames, but we don't know the total size ( $N$ )
- RS algorithms can pick random subsets from a stream of data
  - **Interview** Question
  - Given a stream of numbers, sample 1 number uniformly
    - Harder version: sample  $k$  numbers uniformly

# Importance Sampling

- **Importance** sampling is a statistical technique used primarily to estimate properties of a particular distribution, while only having samples generated from a **different distribution** rather than the distribution of interest
- Applications in Machine Learning
  - Rare Event Simulation: Sample the 'important' parts of a space where rare events occur.
  - Monte Carlo Methods: improve the efficiency of Monte Carlo simulations.
  - Reinforcement Learning: In scenarios like off-policy learning
  - Variational Inference: approximate intractable integrals in Bayesian inference
- Study in the future when you do deeper with probabilistic modeling

*“Acquire knowledge and impart it to the people.”*

*“Seek knowledge from the Cradle to the Grave.”*



