

# Long-Tailed Learning

## Question Setting

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- **Imbalanced Train Set and Balanced Test Set**

## Trends of Solutions

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- **Data level techniques: SAMPLING**
  - **over-sampling**
    - disadvantages: overfitting to less samples
    - less followers
  - **down-sampling**
    - disadvantages: loss of information
    - less followers
  - **class-balanced sampling**
    - combine over-sampling and down-sampling
  - **mixing minor samples**
    - SMOTE: Synthetic Minority Over-sampling Technique
- **Loss level techniques: REWEIGHTING**
  - **loss level:** make minor classes' misclassification cost higher
    - ratio of total population to class population
  - **score level:** balance the predicting scores directly
    - Training Cost-Sensitive Neural Networks with Methods Addressing the Class Imbalance Problem
- **Others**
  - **Transfer based techniques**

- Clustering based techniques

## Popular Benchmarks

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- ImageNet-LT
- Places365-LT
- iNaturalist

## Related Papers

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### Training Cost-Sensitive Neural Networks with Methods Addressing the Class Imbalance Problem

$$O_i^* = \eta \sum_{c=1}^C O_i \text{Cost}[i, c].$$

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- Cost based moving threshold can achieve similar performance with the sampling methods.
- Cost function should be pre-defined.

### SMOTE: Synthetic Minority Over-sampling Technique

$$r = p + (q - p) \times \lambda$$

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- Generate synthetic examples from K-means neighbours by operating in “feature space”
- Combine over-sampling and down-sampling

## Take-on Message

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- For sampling: The performance depends on the class DISCRIMINATORY INFORMATION contained in the PRUNED PORPULATION.
- SCORE LEVEL BALANCING is underdeveloped
- Balancing will hurt feature learning, while help classifier learning