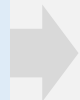


Modalities have imbalanced optimization degree  
(*CVPR 22 ORAL*)



Modality imbalance has sample-level discrepancy  
(*CVPR 24*)



Unimodal and multimodal objectives have conflicts  
(*ICML 24*)



- **Optimization-aware**: Alleviate optimization imbalance via on-the-fly gradient modulation (*CVPR 22 ORAL*) and bi-directional modulation (*T-PAMI 24*); Or considering the intrinsic limitation within modalities, break the reliance on preferred modality by soft re-learning (*ECCV 24*). Or recover suppressed information acquisition with the guidance of Fisher Information Matrix (*CVPR 25*).
- **Data-aware**: Observe fine-grained modality imbalance by Shapley-value based evaluation, and ease such an imbalance by targeted re-sample strategy (*CVPR 24*).
- **Learning objective-aware**: Introduce multimodal Pareto integration to solve conflicts between unimodal and multimodal objectives (*ICML 24*); Analyze the robustness limitation imposed by modality imbalance, and improve multimodal robustness by targeted designed learning objective (*ICLR 24*).

### Exploration in classic multimodal models



Current PEFT for MLLMs often ignore multimodal characteristics.  
(*NeurIPS 25 ORAL*)



MLLMs should consider both unimodal and cross-modal adaptation. A new PEFT method, Moka, is proposed to effectively capture unimodal information and enhance cross-modal interaction (*NeurIPS 25 ORAL*).

### Exploration in MLLMs scenario