

CoAnnotating: Uncertainty-Guided Work Allocation between Human and Large Language Models for Data Annotation

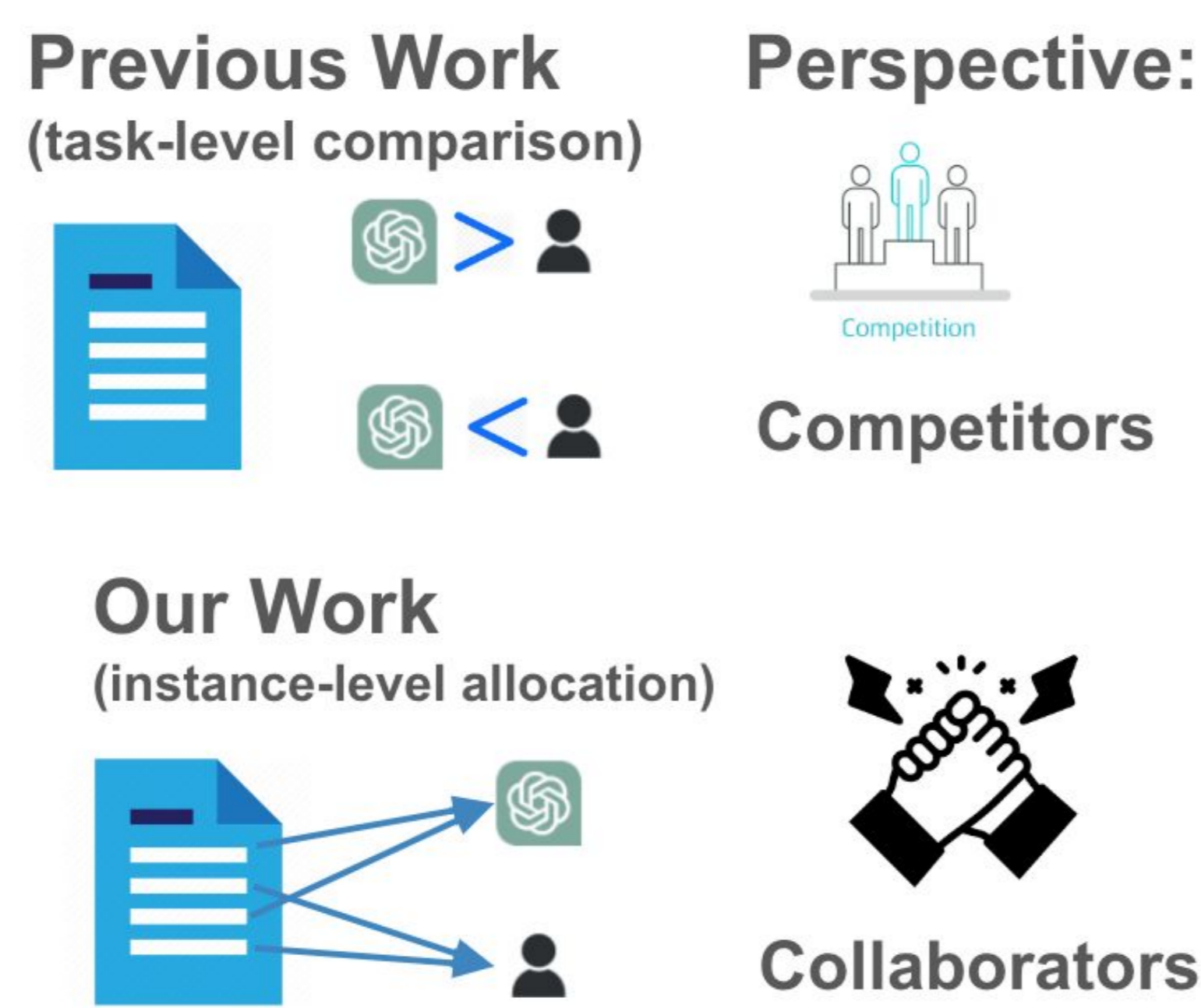
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1. Introduction

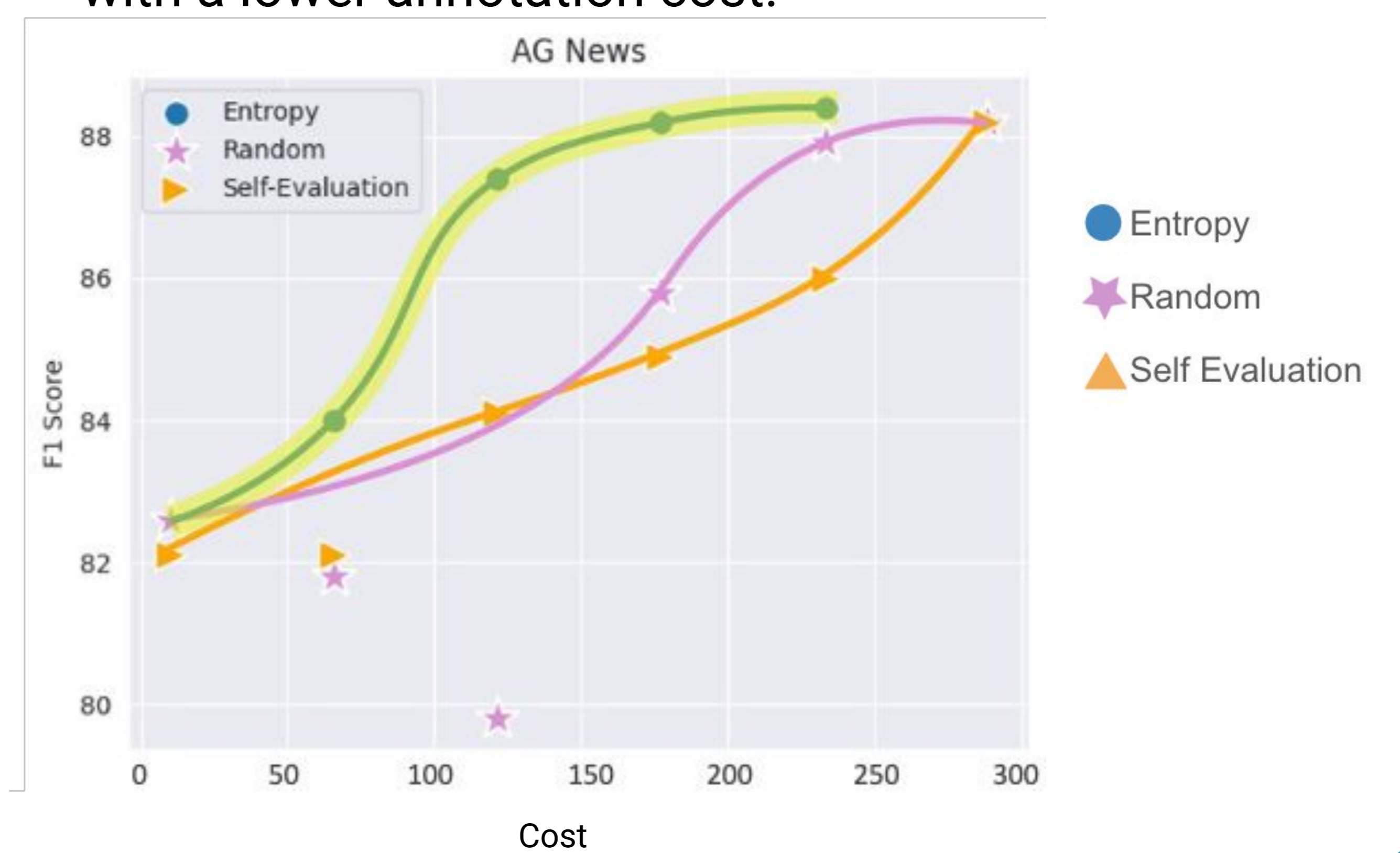
- A novel paradigm for human-Large Language Model (LLM) collaboration for data annotation: **CoAnnotating**



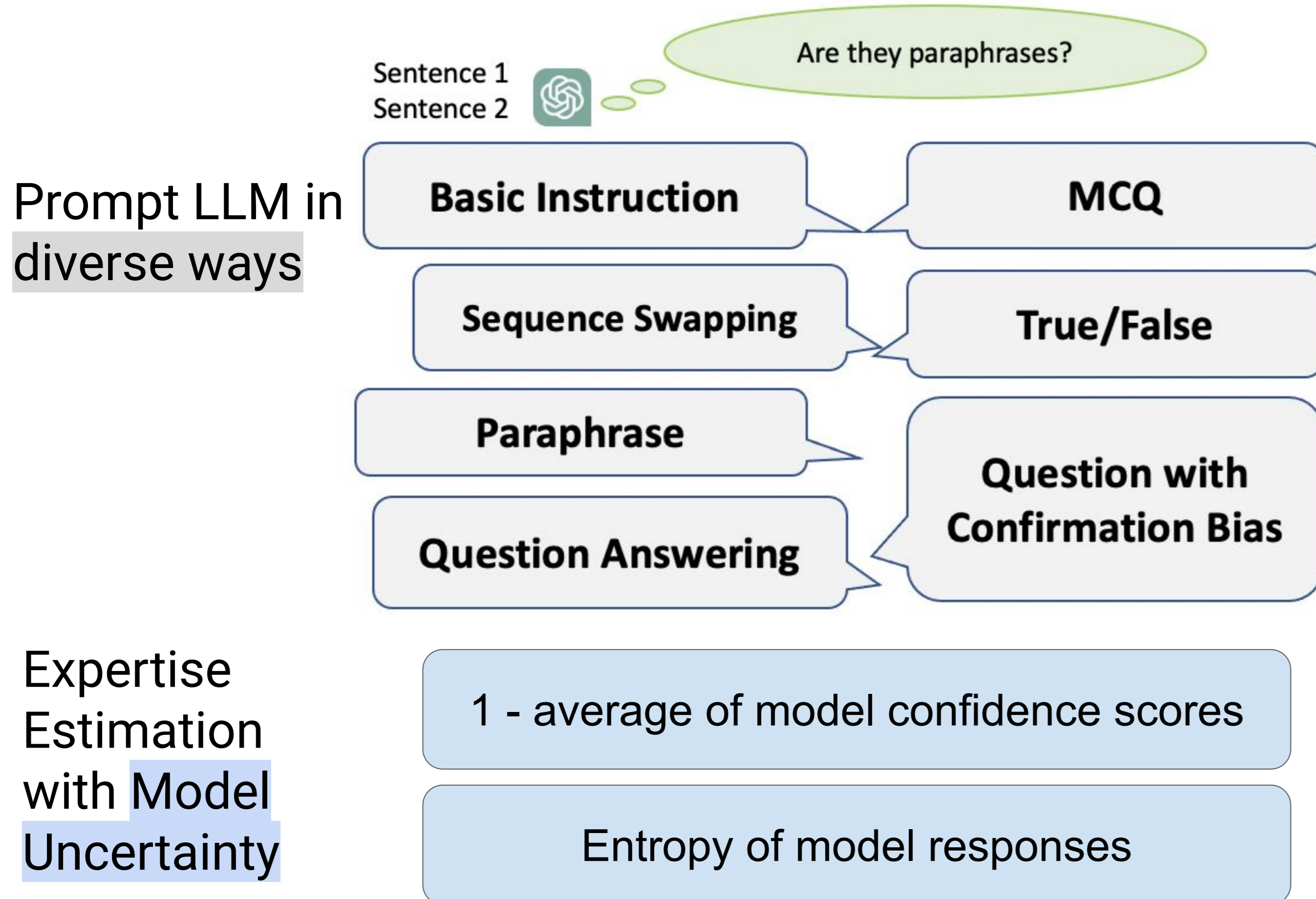
- High Annotation Quality
- Low Annotation Costs

4. Allocation Analysis

- Pareto Efficiency:** an optimal strategy that it is impossible to further improve the annotation quality with a lower annotation cost.

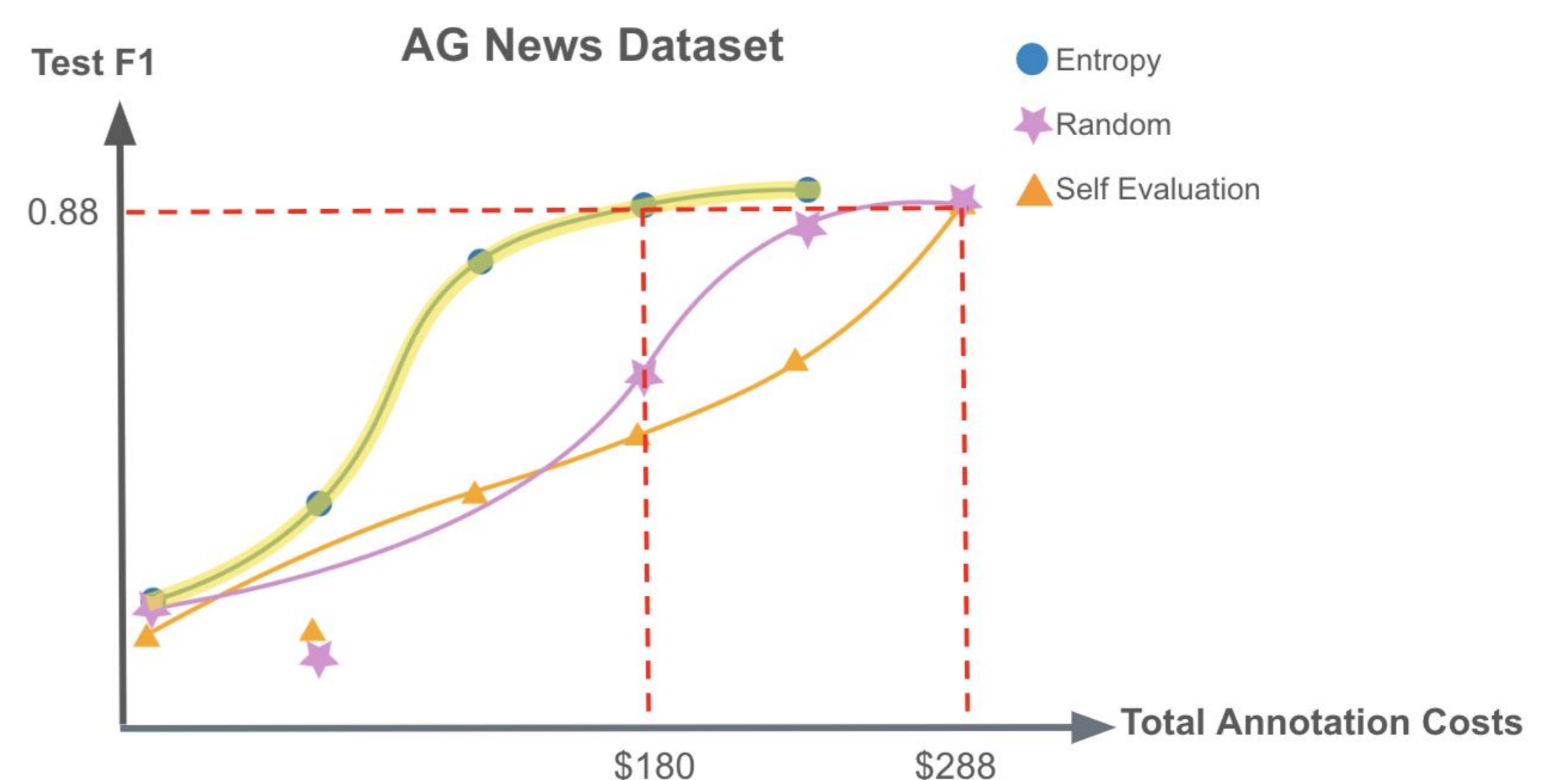


2. Expertise Estimation

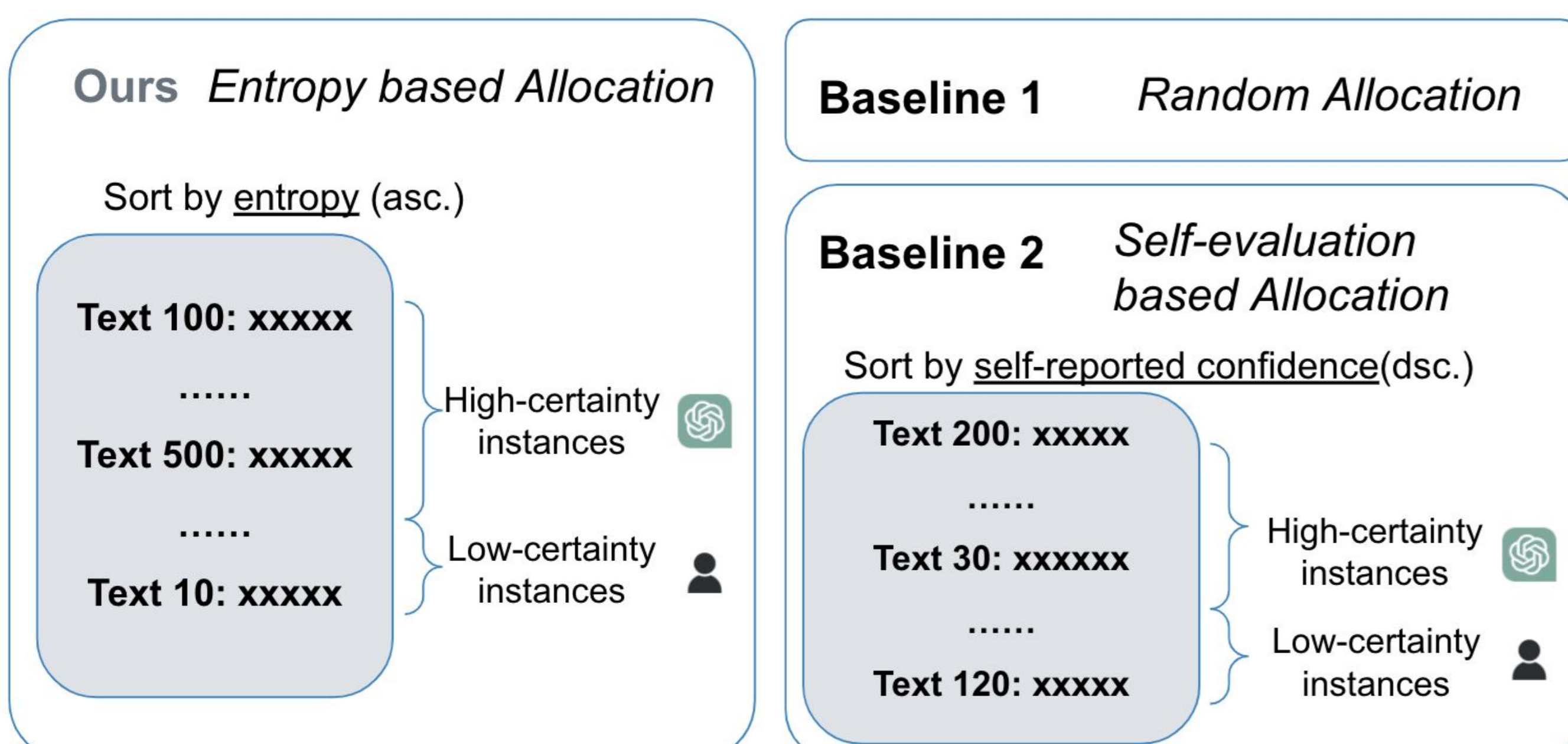


5. Allocative Efficiency

% LLM	0	20	40	60	80	100
Random	57.5	55.9	53.2	46.2	50.3	42.0
Self-Eval	57.5	57.8	55.9	51.8	52.9	43.0
Entropy	57.5	58.4	56.9	55.9	53.8	42.0



3. Work Allocation



6. Conclusion + Future Directions

- LLM Annotation for All Data: Economical, Subhuman-Level Performance (70)
 - Manual Annotation for All Data: Costly, Human-Level Performance (90)
 - Manual Annotation for All Data: Greater Cost Efficiency (90)
- Consider variations within human and LLM annotators; aim for superhuman-level performance