

# Reducing Hallucinations in Abstractive Summarization via Verifier-Reranking

## Proposal

Abstract summarizers often produce fluent but unsupported statements that limit practical use. This project targets factuality for news summarization with a simple, reproducible approach. I will fine-tune BART-base on the public CNN/DailyMail dataset. At inference, I will generate K candidate summaries per article and rerank them using an automatic factuality score. Using **FactCC** as the primary verifier and reporting **QAGS** as a secondary check. If reranking harms ROUGE, I will test constrained decoding that favors copying from the source as a fallback. Implementations rely on Hugging Face Transformers, Datasets, and Evaluate, and will be cited.

## Dataset

CNN/DailyMail (Hugging Face ID: `ccdv/cnn_dailymail`, `config 3.0.0`) with standard train/validation/test splits.

## Metrics and Success

Quality: ROUGE-1/2/L and BERTScore on validation and test.

Factuality: FactCC (primary) and QAGS on validation.

Human check: 50 examples with an error taxonomy.

**Success criterion:**  $\geq +2.0$  FactCC points over the BART baseline on validation with  $\leq 1.0$  ROUGE-L drop.

## Risks / Challenges

Verifier may mis-score paraphrases; mitigate with human spot checks. Compute limits handled by base-size models and modest K. If reranking underperforms, tune decoding and report ablations.

## Why this matters

Abstractive models often sound good but insert mistakes. Picking the most factual draft reduces those hallucinations without heavy engineering.

*This focused design addresses a known failure mode with minimal engineering and clear metrics. It fits the class scope, uses a single public dataset with stable splits, and supports deep analysis. I will release code, configs, and small data samples for reproducibility.*

## References

- **BART (Lewis et al., 2020, ACL)**  
ACL Anthology page: <https://aclanthology.org/2020.acl-main.703/>  
PDF: <https://aclanthology.org/2020.acl-main.703.pdf>
- **PEGASUS (Zhang et al., ICML 2020)**  
ICML / MLR page: <https://proceedings.mlr.press/v119/zhang20ae.html>  
PDF: <https://proceedings.mlr.press/v119/zhang20ae/zhang20ae.pdf>  
Project page: <https://jingqingz.github.io/publication/2019-PEGASUS>
- **FactCC (Kryściński et al., EMNLP 2020)**  
Anthology (paper “Evaluating the Factual Consistency of Abstractive Text Summarization”): <https://aclanthology.org/2020.emnlp-main.750/>  
PDF: <https://aclanthology.org/2020.emnlp-main.750.pdf>
- **QAGS (Wang et al., ACL 2020)**  
ACL Anthology page: <https://aclanthology.org/2020.acl-main.450/>  
PDF: <https://aclanthology.org/2020.acl-main.450.pdf>
- **BERT (Devlin et al., NAACL-HLT 2019)**  
ArXiv: <https://arxiv.org/abs/1810.04805>  
Conference reference: Jacob Devlin, Ming-Wei Chang, Kenton Lee, Kristina Toutanova.  
*BERT: Pre-training of Deep Bidirectional Transformers for Language Understanding*.  
NAACL-HLT 2019.