Automated Generation of Hospital Discharge Summaries Using Clinical Guidelines and LLMs

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Results

Section

Admission Details Allergies And Adverse Reaction Clinical Summary Diagnoses **Discharge Details** Patient Demographics Plan And Requested Actions Social Context

> Macro Average Micro Average

3 Health Data Research UK, London, UK 4 University College London Hospitals NHS Foundation Trust, London, UK

Recall	Precision	Acc
0.90	0.95	0.85
0.98	1.00	0.98
0.76	0.92	0.71
0.84	0.94	0.80
0.93	0.96	0.89
1.00	0.84	0.84
0.90	0.88	0.80
0.96	0.88	0.84
0.91	0.92	0.84
0.86	0.92	0.81

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Findings

- Clinical guidelines make good prompts - Few shot learning overcomes training data scarcity - LLMs are good at manipulating language but poor at making clinical judgements

Path to Clinical Adoption

Improved evaluation

- Data sources
- Clinical framework (low annotator agreement 59.72%)
- Automated evaluation for methodological hill climbing

Data governance-compliant LLM

Workflow integration

- Human in the loop -> automation bias
- Handling of error cases
- EHR Integration



Paper



Code







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