

FedMentor: Domain-Aware Differential Privacy for Heterogeneous Federated LLMs in Mental Health



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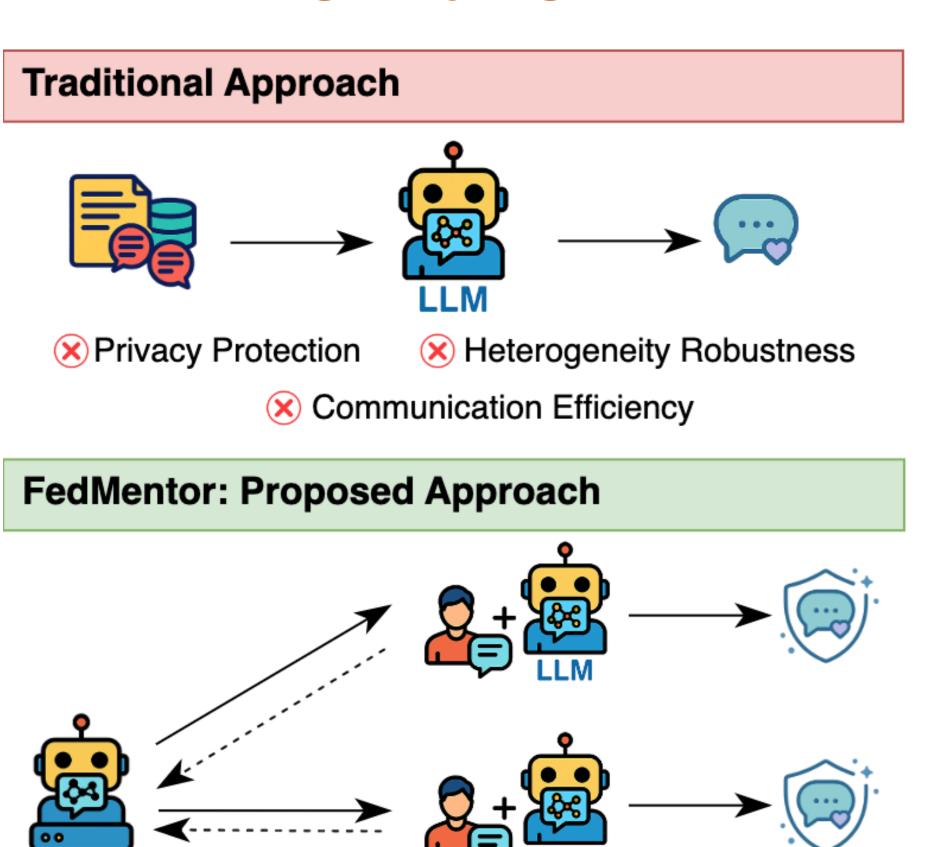


TL;DR

FedMentor combines domain-aware DP with efficient LoRA training, matching FL utility and improving safety across heterogeneous mental health datasets.



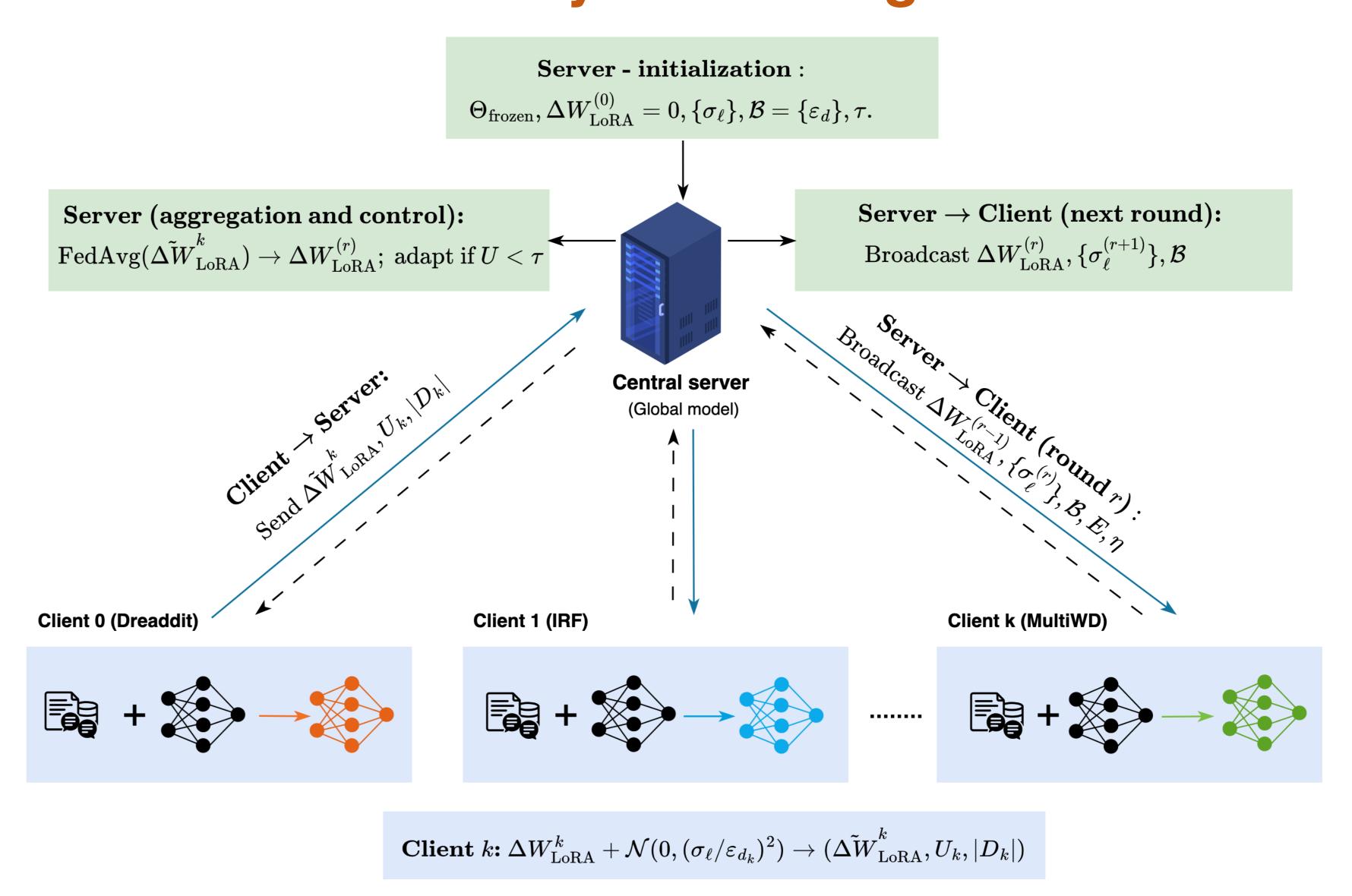
Motivation



✓ LoRA-Only Efficiency

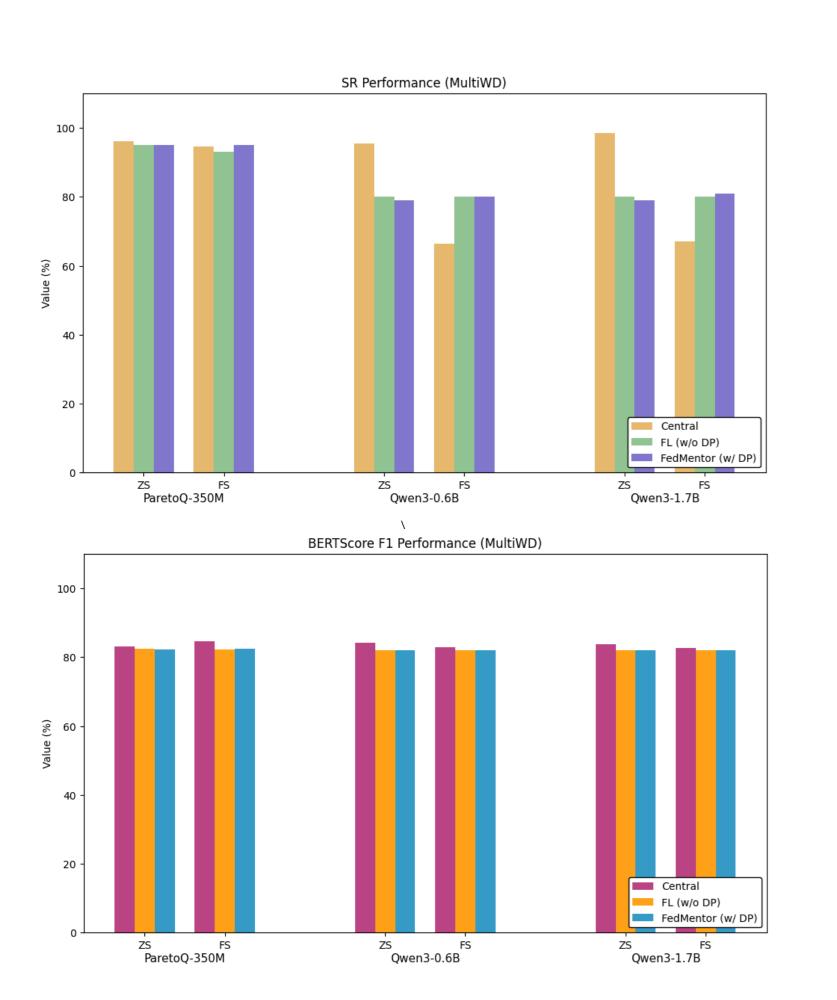
Non-IID Robustness

FedMentor: Privacy-Preserving Federated LLMs

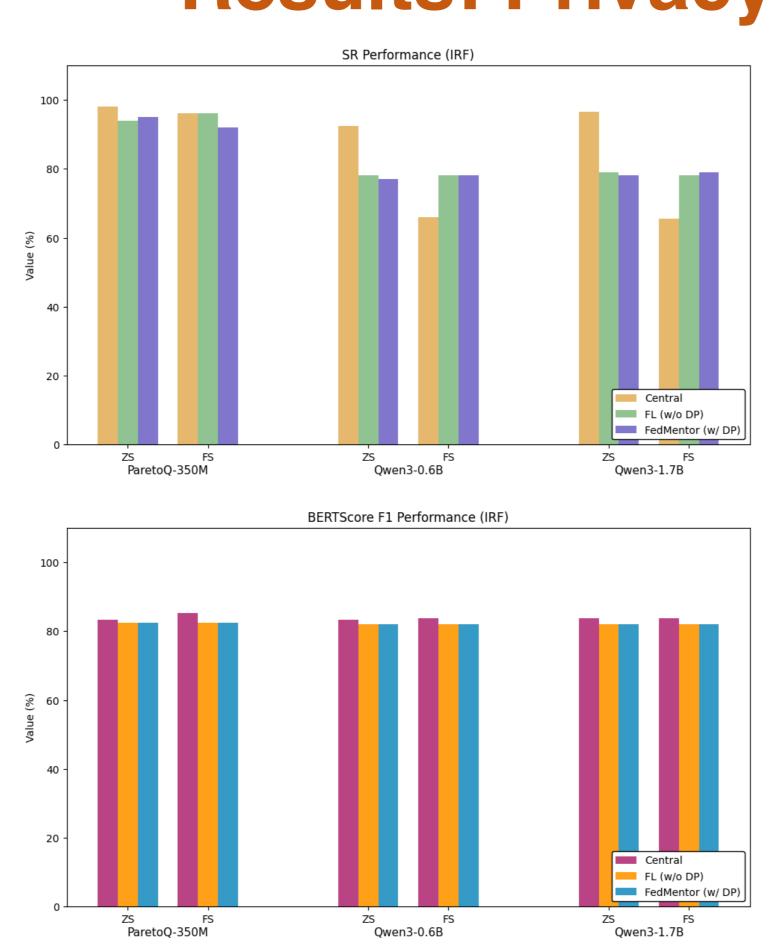


Terminate / Output: After R rounds or budget exhaustion: $\mathcal{M}^{(R)} = (\Theta_{\mathrm{frozen}}, \Delta W_{\mathrm{LoRA}}^{(R)})$

Results: Privacy & Safety



Domain-Aware Privacy



- FedMentor matches FL utility; privacy cost is minimal, with safety and quality aligned
- **Safe rates** stay near FL across Dreaddit, IRF, MultiWD and reach **95%** on MultiWD
- Across Dreaddit, IRF, MultiWD, B-F1 stays within **0.2 percentage** points of FL

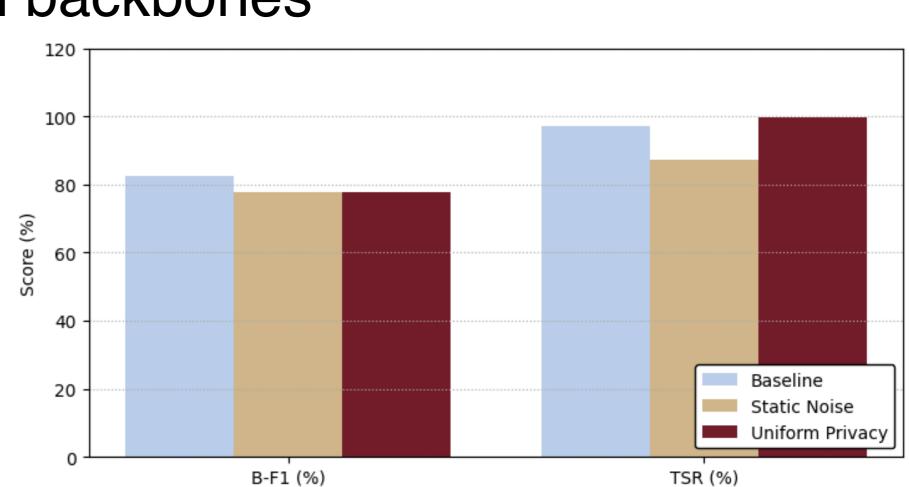
Results: Practical Efficiency

- Adapter updates stay small at 17–67 MB; communication
 50–173 MB per round
- Peak memory fits 80 GB, with ~34 GB (ParetoQ-350M) and ~78 GB (Qwen3-0.6B/1.7B) footprints

Model	Adpt (MB)	Comm (MB)	Mem (GB)	Time (min)
Global summary (aggregated)				
ParetoQ-350M	16.56	49.69	33.74	7.64
Qwen3-0.6B	38.50	110.00	77.86	10.78
Qwen3-1.7B	66.50	172.90	77.86	11.37
Per-dataset breakdown (Client centric)				
MultiWD				
ParetoQ-350M	16.56	49.69	33.74	7.64
Qwen3-0.6B	38.50	110.00	76.70	11.37
Qwen3-1.7B	66.50	172.90	77.86	12.25

FedMentor Stays Robust

- Uniform privacy maintains TSR ~ 99–100%
- Across ε 0.1–1.0, B-F1 holds 78–82% range for all backbones



Conclusions

- We introduced FedMentor with domain-aware DP, safe rate 95%, B-F1 within 0.2% difference under heterogeneous clients.
- LoRA-only updates send 50–173 MB per round and fit 80 GB GPUs, while uniform privacy stabilizes safety across models and datasets.

