

Initial Insights Into a Secure Large Language Model for MRI Examination Requests: A Retrospective Study

James Thomas Patrick Decourcy Hallinan^{1,2}, MBChB; Naomi Wenxin Leow³, BComp, MComp; Yi Xian Low¹, MBBS; Aric Lee¹, MBBS; Wilson Ong¹, MBBS; Matthew Ding Zhou Chan¹, BmedSc, MD; Gordan Toh Cheong Zheng⁵, Diploma in Biomedical Science; Ganakirthana Kalpenya Devi¹, MbBchBao; Stephanie Shengjie He¹, MBBS; Daniel De-Liang Loh¹, MBBS, MRCS; Desmond Shi Wei Lim¹, MBBS; Xi Zhen Low¹, MBBS; Mei Chin Lim¹, MBBS; Clement Yong¹, MBBS; Weizhong Jonathan Sng¹, MBBS; Ee Chin Teo¹, MMRT; Jiong Hao Tan⁴, MBBS, MMed, MRCS; Naresh Kumar⁴, MBBS, MS, DNB, DM; Andrew Makmur^{1,3}, BmedSc, MBBS, MMed; Yonghan Ting¹, MBBS

¹ Dept. of Diagnostic Imaging, NUH, Singapore

² Dept. of Diagnostic Radiology, YLL School of Medicine, NUS, Singapore

³ AI Office, NUHS, Singapore

⁴ National University Spine Institute, NUH, Singapore

⁵ Republic Polytechnic, Singapore

Correspondence:
Dr James Hallinan, Senior Consultant, Department of Diagnostic Imaging, National University Hospital
Email: james_Hallinan@nuhs.edu.sg

Background

- Incomplete Magnetic Resonance Imaging (MRI) requests lead to suboptimal protocols, repeat studies & reduced accuracy.
- LLMs can interpret clinical text at scale, but privacy constraints limit direct clinical use
- Institutional sLLMs operate behind firewalls, protecting patient data while enabling augmentation

Objectives

For all MRI subspecialty modalities (Musculoskeletal, Neuroradiology, body):

1. Does sLLM augmentation improve clinical information quality of MRI examination requests?
2. Can sLLM match radiologist-level MRI protocol accuracy?

Materials and Methods

Figure 1: Summary of materials

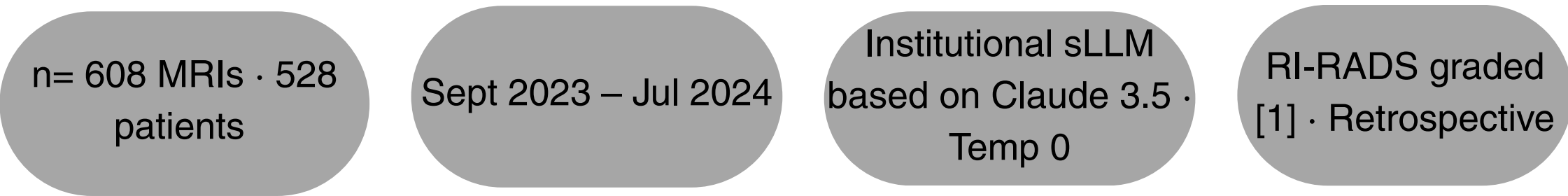


Figure 2: Flowchart of sLLM augmentation

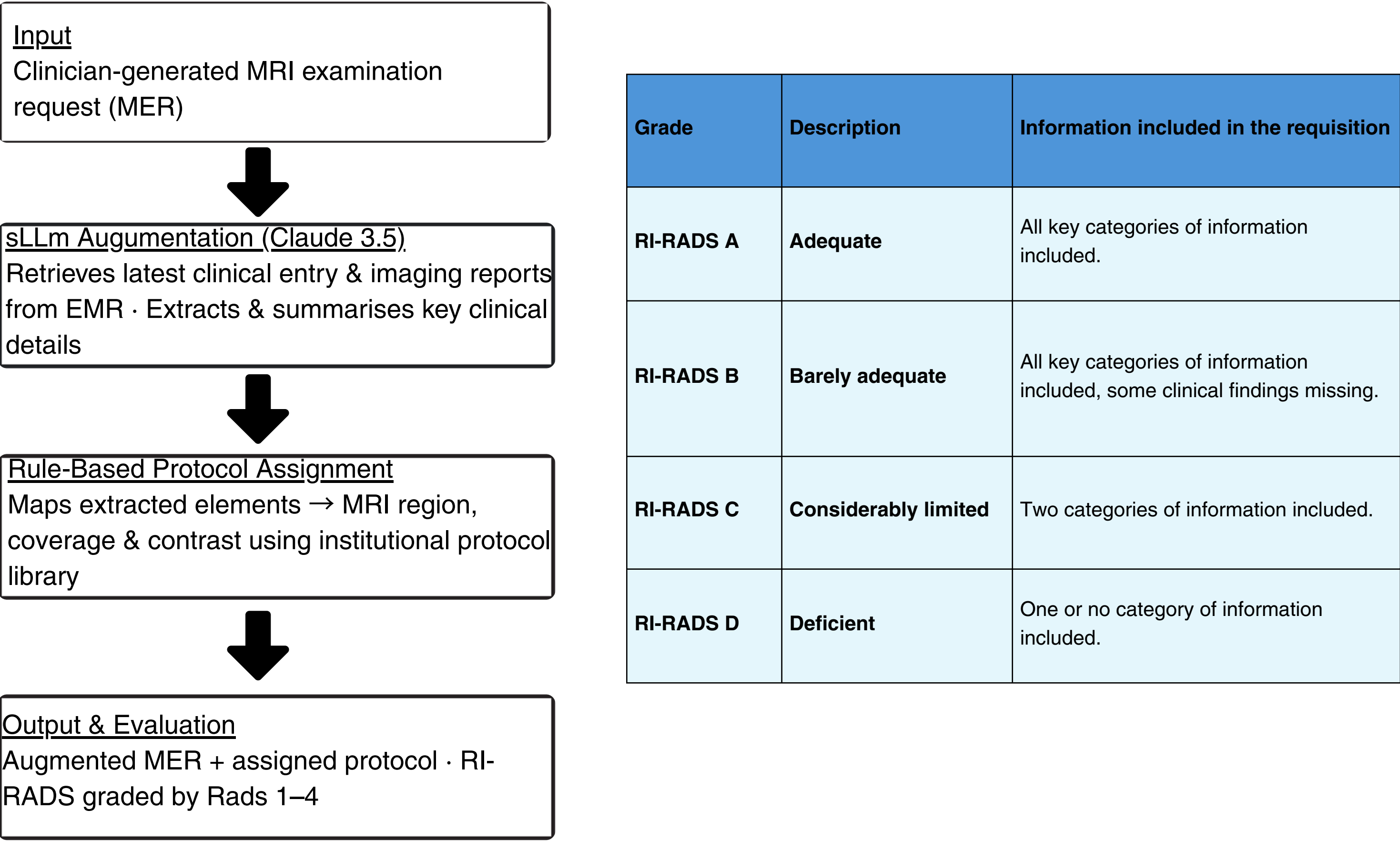
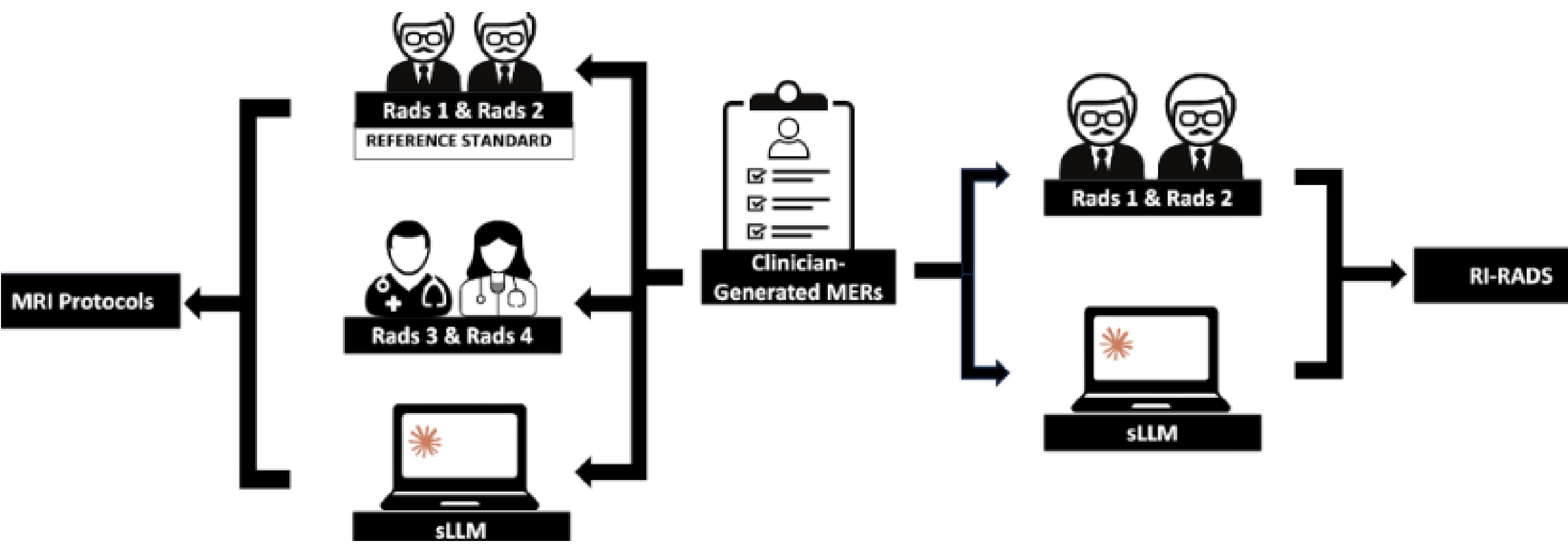
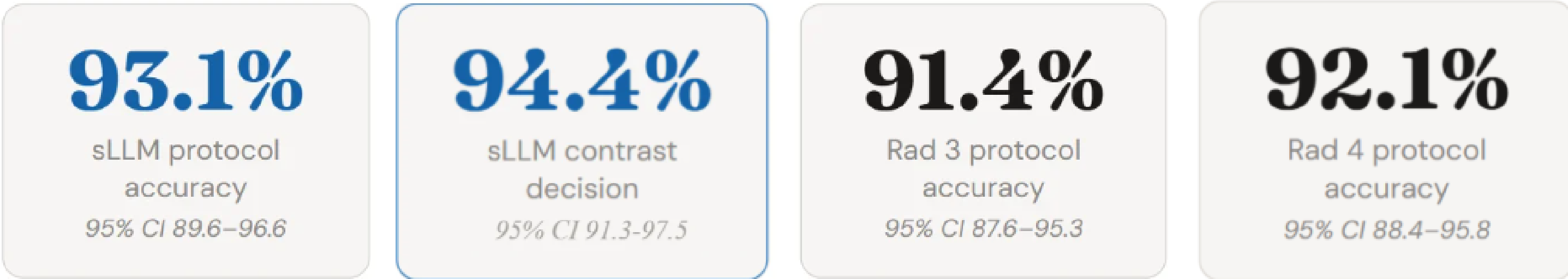
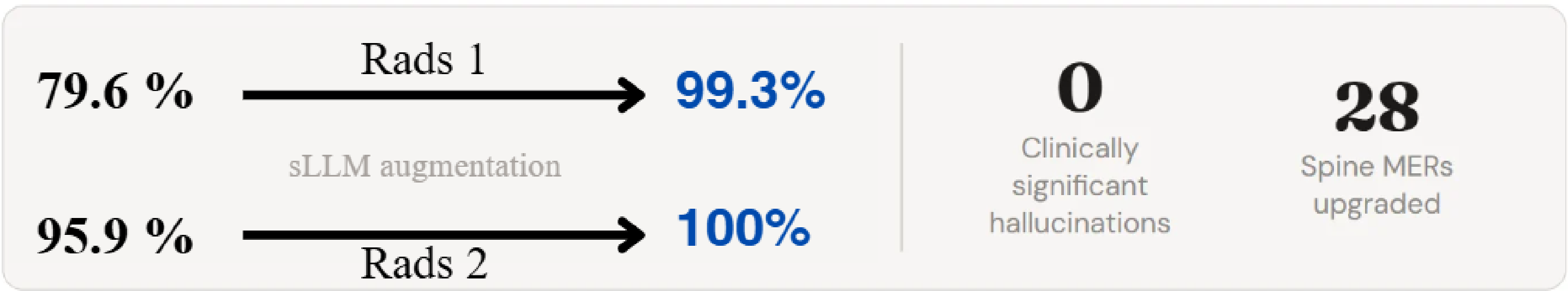


Figure 3: Study design flowchart. Clinician MERs were augmented using an sLLM (Claude 3.5, Anthropic). Two experienced radiologists (Rads 1 and 2) graded information quality via RI-RADS and provided the protocol accuracy reference standard. The sLLM and two general radiologists (Rads 3 and 4) were evaluated against this standard.



Main Findings

Figure 4: Summary of results of MRI protocolling accuracy for the sLLM and board-certified radiologists versus reference standard. sLLM augmentation nearly eliminated inadequate MRI requests and matched expert radiologist protocol accuracy.



ACI = 0.97 (almost perfect interobserver agreement) · Consensus protocol in 93.1% of cases

Figure 5: Overall MRI Protocolling accuracy for the sLLM and board-certified radiologists versus reference standard.

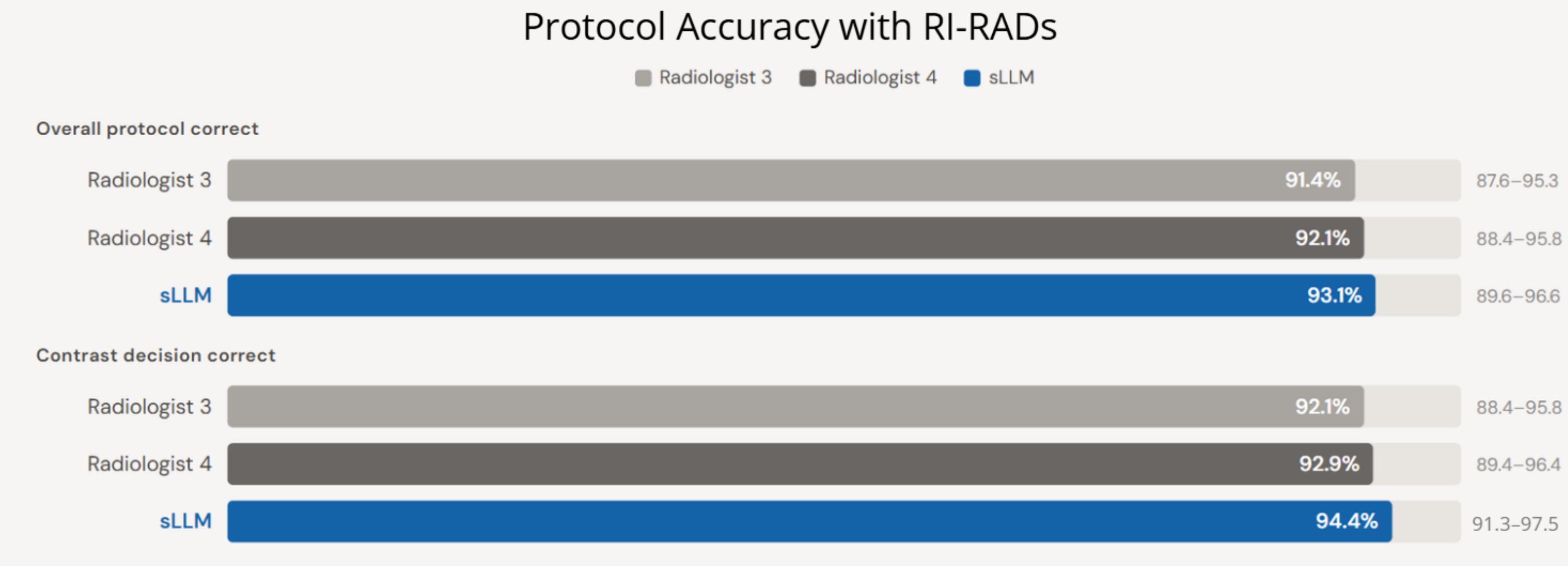
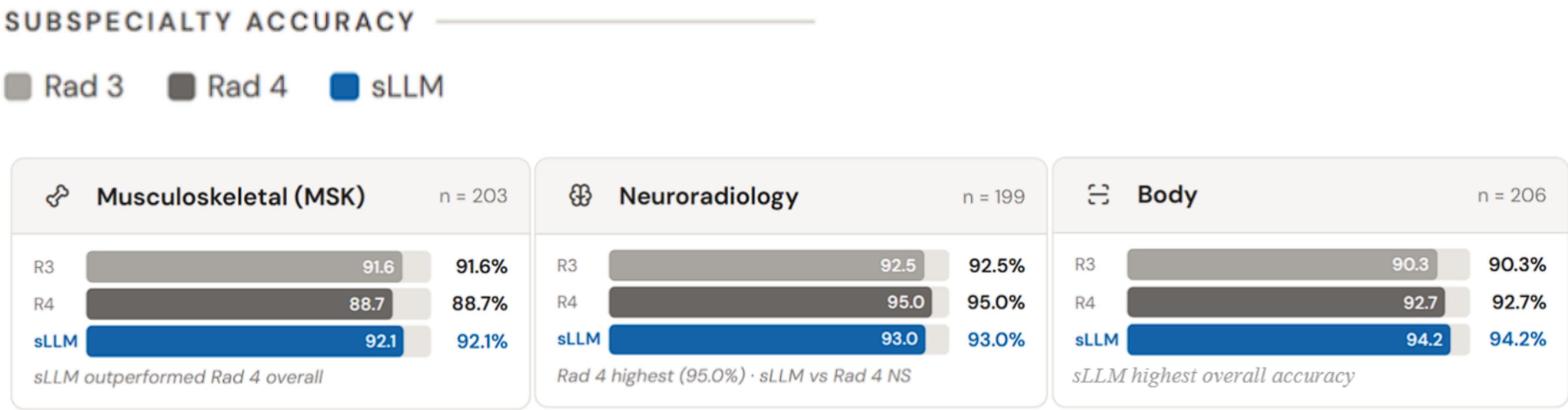


Figure 6: MRI protocolling accuracy for the sLLM and board-certified radiologists versus reference standard, broken down into subspecialties



Discussion

- sLLM upgraded MER quality across all subspecialties, fewer than 1% remained inadequate post-augmentation
- sLLM protocol accuracy (93.1%) was comparable to board-certified radiologists (Rad 3: 91.4%, Rad 4: 92.1%); consistent with prior LLM studies in spine/MSK [2] and CT protocol assignment [3].
- Critical clinical details absent from original MERs (surgery history, tumour, infections) were extracted and summarised by the sLLM
- Rule-based script made final protocol decisions — sLLM provided clinical context only, not the decision

Limitations

- Single-centre, retrospective study · outpatient MERs only
- Not powered for non-inferiority testing
- Post-deployment guardrails remain essential

Conclusion

- Near 100% adequate MERs after sLLM augmentation vs 79.6-95.9% clinician-only
- Protocol accuracy matched experienced radiologists across body, MSK & neuroradiology
- End-to-end pipeline offers a scalable, standardised protocolling pathway with reduced admin burden
- Future work: prospective multicentre evaluation, cost-effectiveness & extension to CT / ultrasound

References

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2. Hitzert MM, et al. Large Language Models for the Automated Vetting of Musculoskeletal MRI Requests. J Digit Imaging. 2024.
3. Kanemaru N, et al. Efficacy of Fine-Tuned Large Language Model in CT Protocol Assignment as Clinical Decision-Supporting System. J Imaging Inform Med. 2025.