

A Perspective on Clinical Deployment of AI-GMA: From Quality-Gated Video Acquisition to Robust Decision Support

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Abstract

GMA is a powerful tool for early cerebral palsy screening, but clinically useful AI-GMA systems must jointly address quality-gated acquisition, robust temporal representation, calibrated risk output, and workflow integration.

- Early identification improves developmental outcomes.
- Quality is the upstream enabler, not the result of fate.
- GMA reduces content bias in the risk.
- Prior AI-GMA work often emphasizes isolated model metrics.
- The goal is trustworthy clinical decision support.

Conventional vs. AI-GMA vs. Deployment-Centered AI-GMA

	Conventional (Heuristic / Rule-based)	Clinical / Workflow Deployment Focus	Deployment-Centered AI-GMA
Input Assumption	Standardized clinical devices	Heterogeneous clinic/home videos with acquisition variability	Heterogeneous clinical forms with quality gating
Primary Target	Risk or AUC on curated sets	PGC / AUC on variability, workflow integration, clinician trust	PGC/AUC on workflow fidelity, integration, clinician trust
Output	Predicted label	Risk score, confidence, uncertainty, escalation path	PGC / AUC on veracity, workflow fidelity, integration, clinician trust
Compute & Privacy	BSN/Transformer (black-box)	3D or privacy-preserving scalable compute requirements	3D or privacy-preserving, scalable compute requirements
Fallure Handling	Continue inference by default	Hold/flag low-quality and edge-case videos	Hold/flag low-quality, explain, escalate, and route to support team

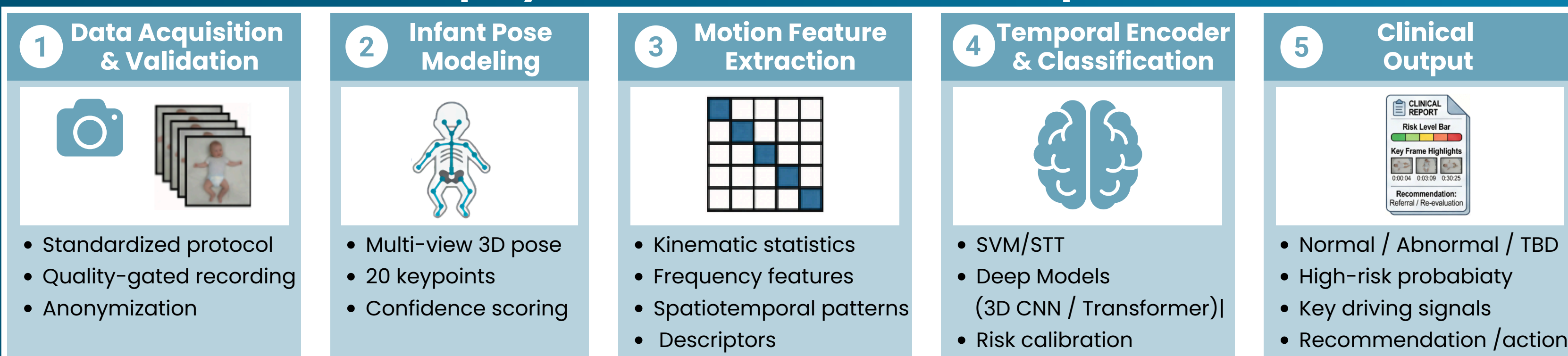
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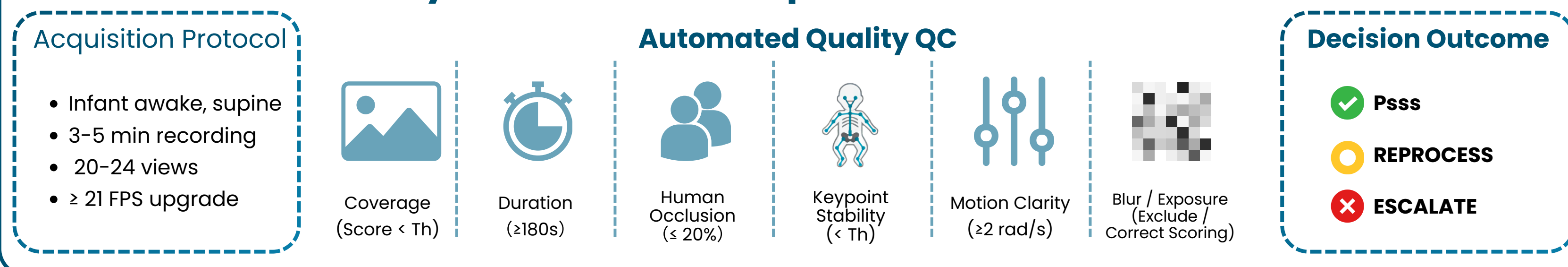
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A call to action for the computer vision community: build infant-adapted models, robust datasets, and privacy-preserving pipelines for trustworthy clinical decision support.

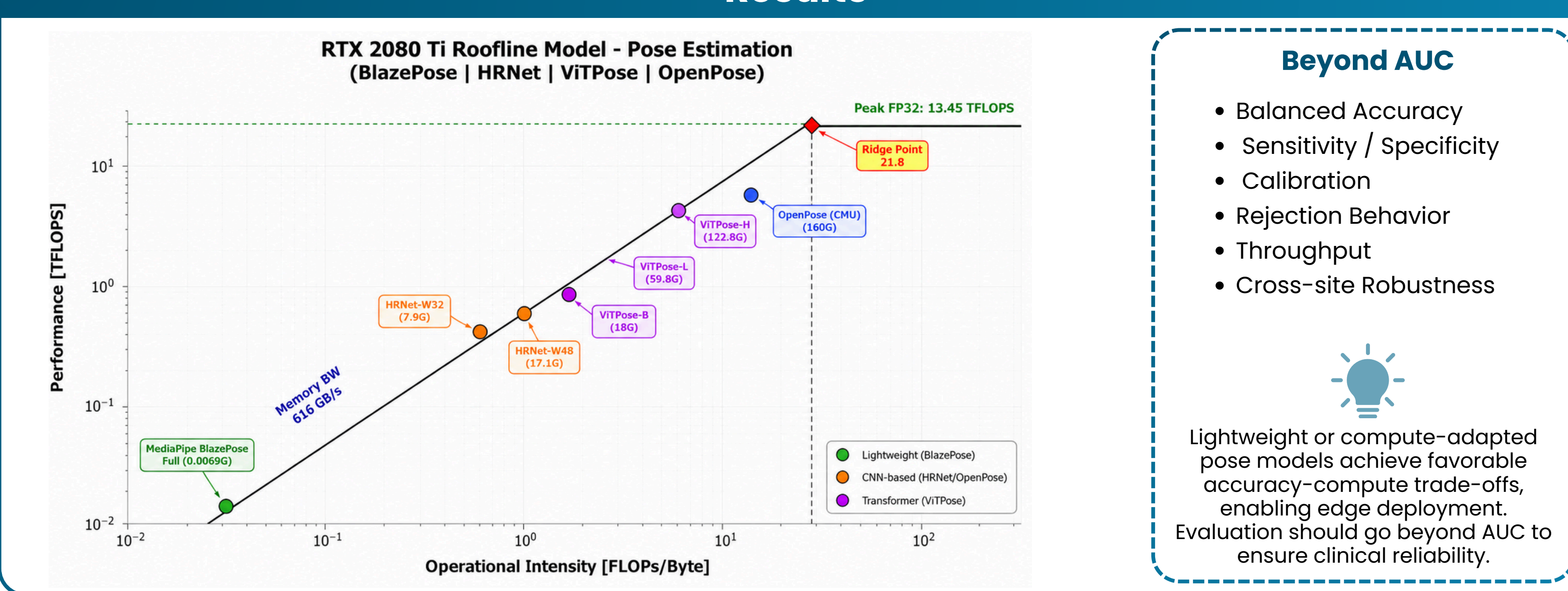
Deployment-Centered AI-GMA Pipeline



Quality-Gated Video Acquisition & QC Framework



Results



Beyond AUC

- Balanced Accuracy
- Sensitivity / Specificity
- Calibration
- Rejection Behavior
- Throughput
- Cross-site Robustness

Lightweight or compute-adapted pose models achieve favorable accuracy-compute trade-offs, enabling edge deployment. Evaluation should go beyond AUC to ensure clinical reliability.

Discussion

- Quality gating before inference is essential for reliability and safety.
- Privacy-preserving compact representations enable safe cross-site collaboration.
- Risk-stratified outputs with calibration and rejection handling support clinicians.
- Seamless integration with pediatric workflows is key for adoption.

Privacy-Preserving Deployment Architecture



Conclusions

- Deployment is the core scientific challenge, not an afterthought.
- Quality gating and QC drive a more reliable and clinically actionable AI-GMA.
- Privacy-preserving collaboration enables broader clinical adoption.
- The goal is trustworthy clinical support systems for early CP screening.

References

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