

Artificial Intelligence for Refractive Error Prediction in Children

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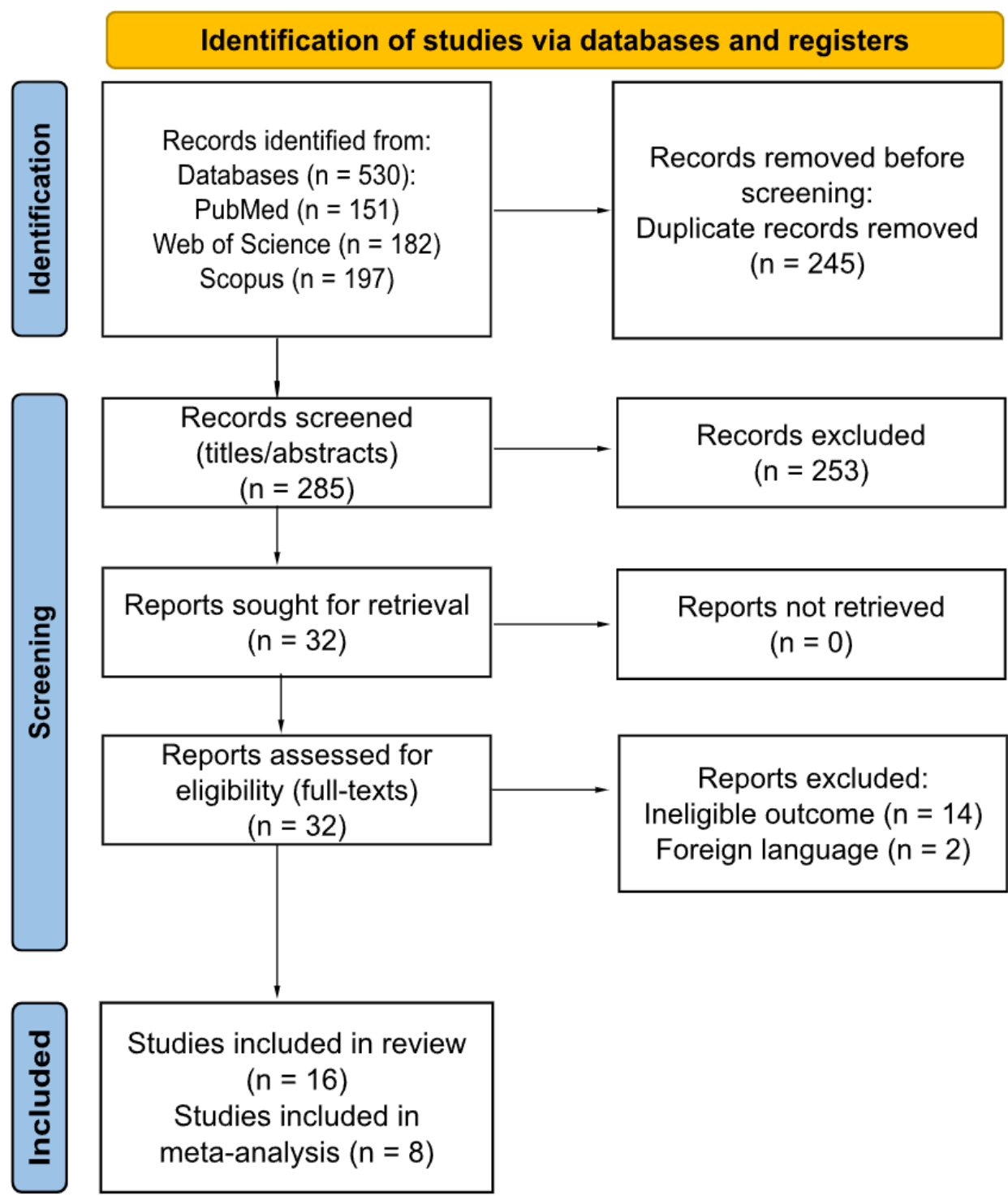
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PURPOSE

To evaluate the performance of artificial intelligence (AI) models in predicting pediatric refractive error from clinical, non-image data.

METHODS

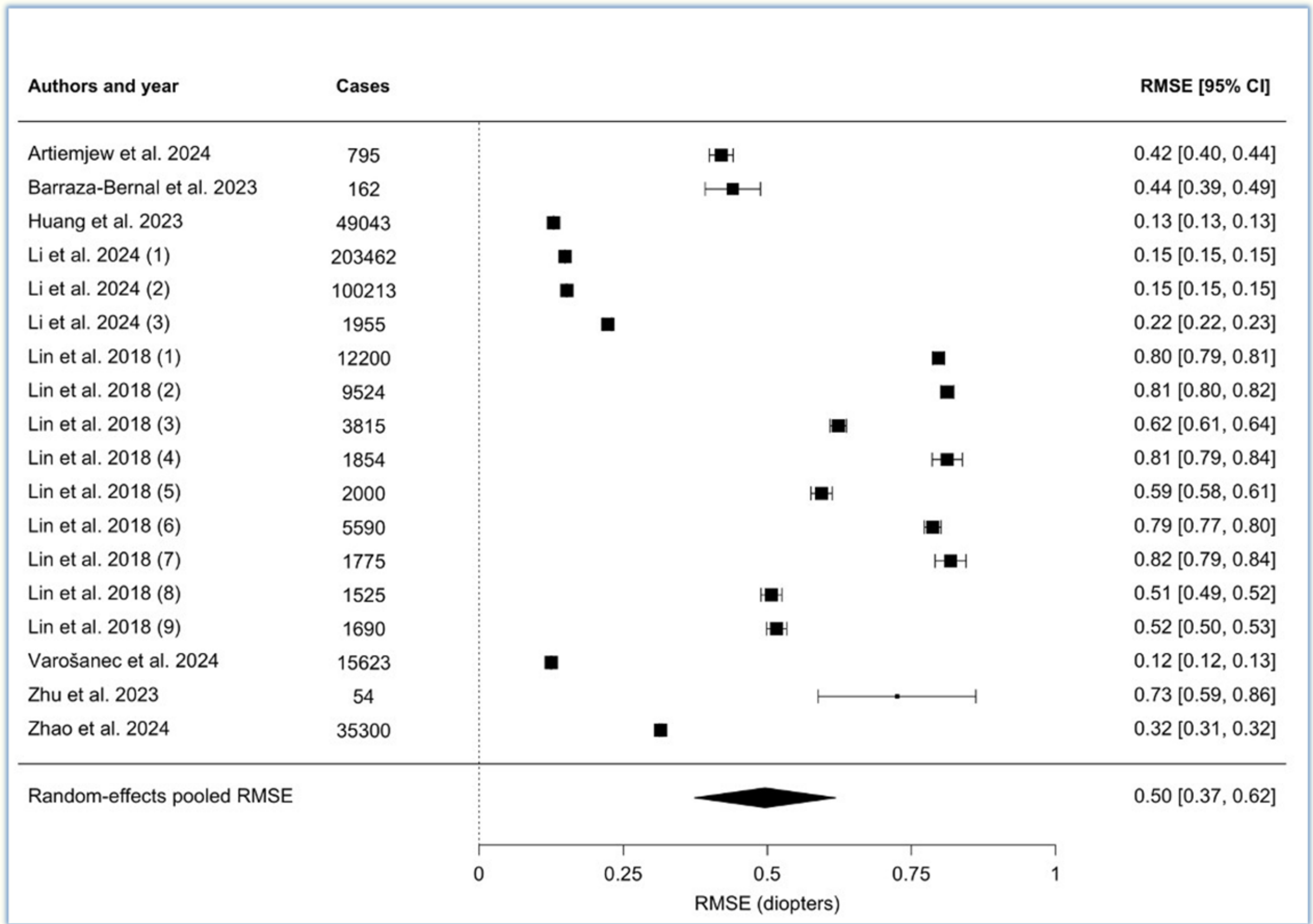
- Systematic review and meta-analysis
- PROSPERO: CRD420251075160
- **Data sources:**
 - MEDLINE, Web of Science, Scopus (April 2025)
- **Eligibility criteria:**
 - Children ≤ 18 years
 - AI-based prediction models
 - Clinical, non-image data
- **Outcomes:**
 - Continuous spherical equivalent (SE) prediction
 - Qualitative synthesis
- **Meta-analysis:**
 - Random-effects model on root-mean-squared error (RMSE)
- **Heterogeneity:**
 - τ^2 , I^2 ; leave-one-out sensitivity analysis
- **Subgroups:**
 - External validation
 - Cycloplegic vs non-cycloplegic refraction
- **Risk of bias:**
 - PROBAST+AI tool



PRISMA Flow Chart

RESULTS

- **Study selection**
 - 16 studies included; 8 in meta-analysis (18 evaluations)
- **Study characteristics**
 - Majority focused on myopia
 - Common inputs: age (69%), sex (56%), parental myopia (50%), axial length (44%), baseline SE (44%)
 - Wide variability in AI models (machine learning, deep learning, ensembles)
 - External validation in 31% of studies
 - Explainability used in 44%
- **Risk of Bias**
 - Generally low risk in participants, predictors, outcomes
 - High/unclear risk in the analysis domain (majority of studies)
- **Meta-analysis**
 - Pooled RMSE: 0.50 D (95% CI: 0.37–0.62)
 - High Heterogeneity: $I^2 = 100\%$
- **Sensitivity analysis**
 - Leave-one-out RMSE: 0.48–0.52 D
- **Subgroup findings**
 - External validation: higher RSME (0.57 vs 0.35 D, $p = 0.092$)
 - Cycloplegic refraction: higher RSME (0.62 vs 0.30 D, $p = 0.002$)



Forest plot of root-mean-squared error (RMSE) for spherical equivalent (SE) refractive error prediction across 18 model evaluations.

CONCLUSION

- ✓ AI models show **moderate accuracy** in predicting pediatric refractive errors from clinical, non-image data.
- ✓ **High heterogeneity limits generalizability.**
- ✓ **Performance decreases** with external validations and cycloplegic refraction.
- **Standardization and robust external validation** are needed.

KEY MESSAGE

- ❖ AI can support **early risk stratification** of pediatric refractive error, but **variability and methodological limitations** currently limit real-world use.