

Objective

- The shortage of neurologists worldwide evokes a need for more efficient diagnostic & patient monitoring measures for neurological diseases
- Goal is to apply machine learning to understand relationships between different structural and functional visual outcomes
- Derive single, optimized measurement of vision** that can eventually be used in self-administered smartphone test

Introduction

- Optical coherence tomography (OCT): structural integrity of the retina focusing on layer of axons and ganglion cells
- Visual evoked potentials (VEP): speed of electrical conduction from retina to visual cortex – prolonged with demyelination
- Brain parenchymal fraction (BPFr) and ventricular CSF volume: measures of global brain atrophy
- Vision is measured by visual acuity (VA) collected under different contrasts. Visual disability based on VA at 1.25% contrast used in multiple sclerosis (MS) clinical trials
- We hypothesize that fitting a linear model to these measurements will decrease measurement error; the area under the curve (AUC) will capture all contrasts, have greater dynamic range & less ceiling effect

Methods

- Develop linear model of VA at different contrasts in healthy donors
- Calculate VA AUC from regression of all VA contrasts
- Validate AUC measure through ICC analysis and correlations with existing clinical outcomes
- Use elastic net regression to generate a model for VA AUC using OCT, VEP, and atrophy data

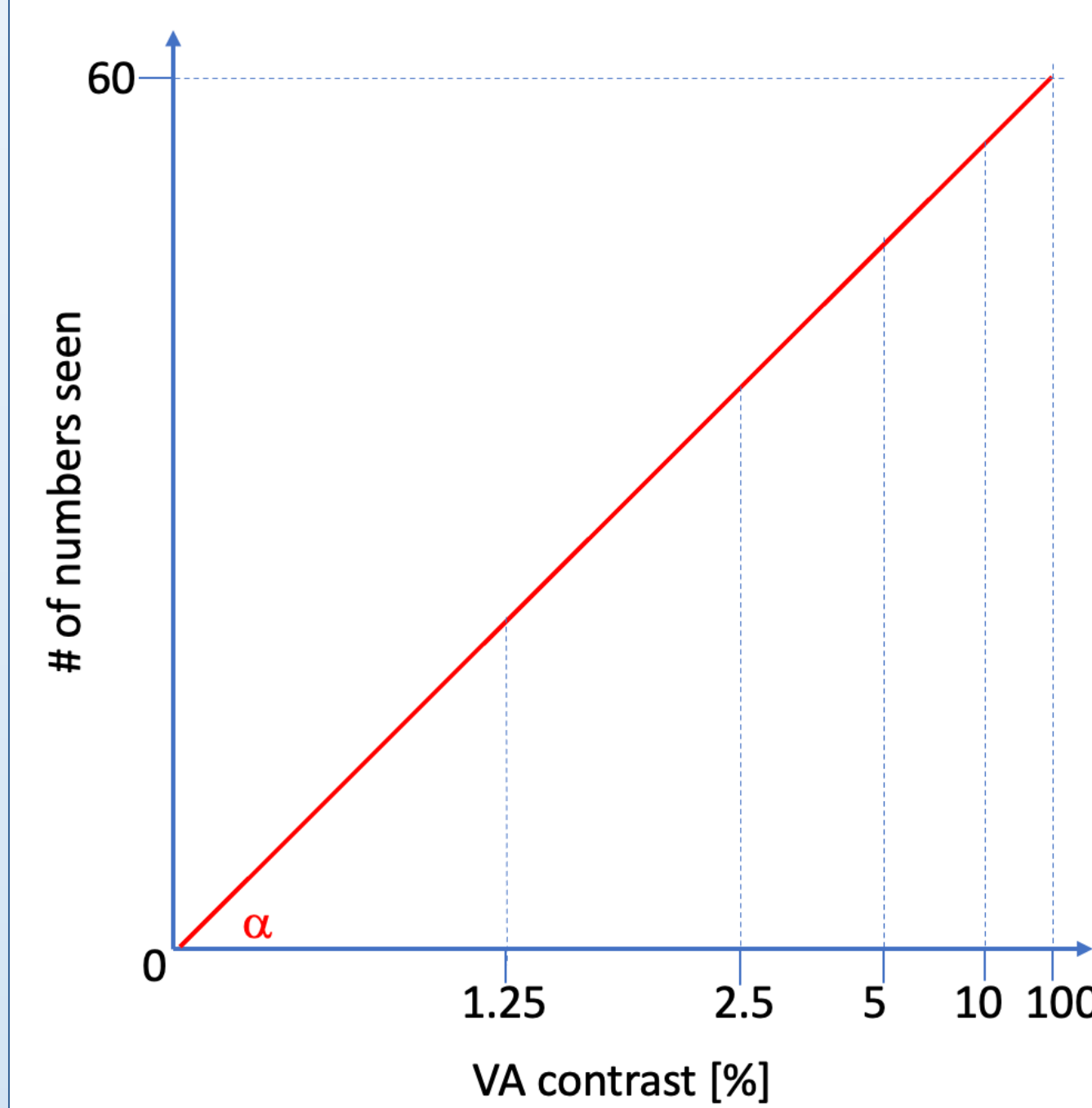


Figure 1: Theoretical linear regression model of VA contrasts to obtain AUC

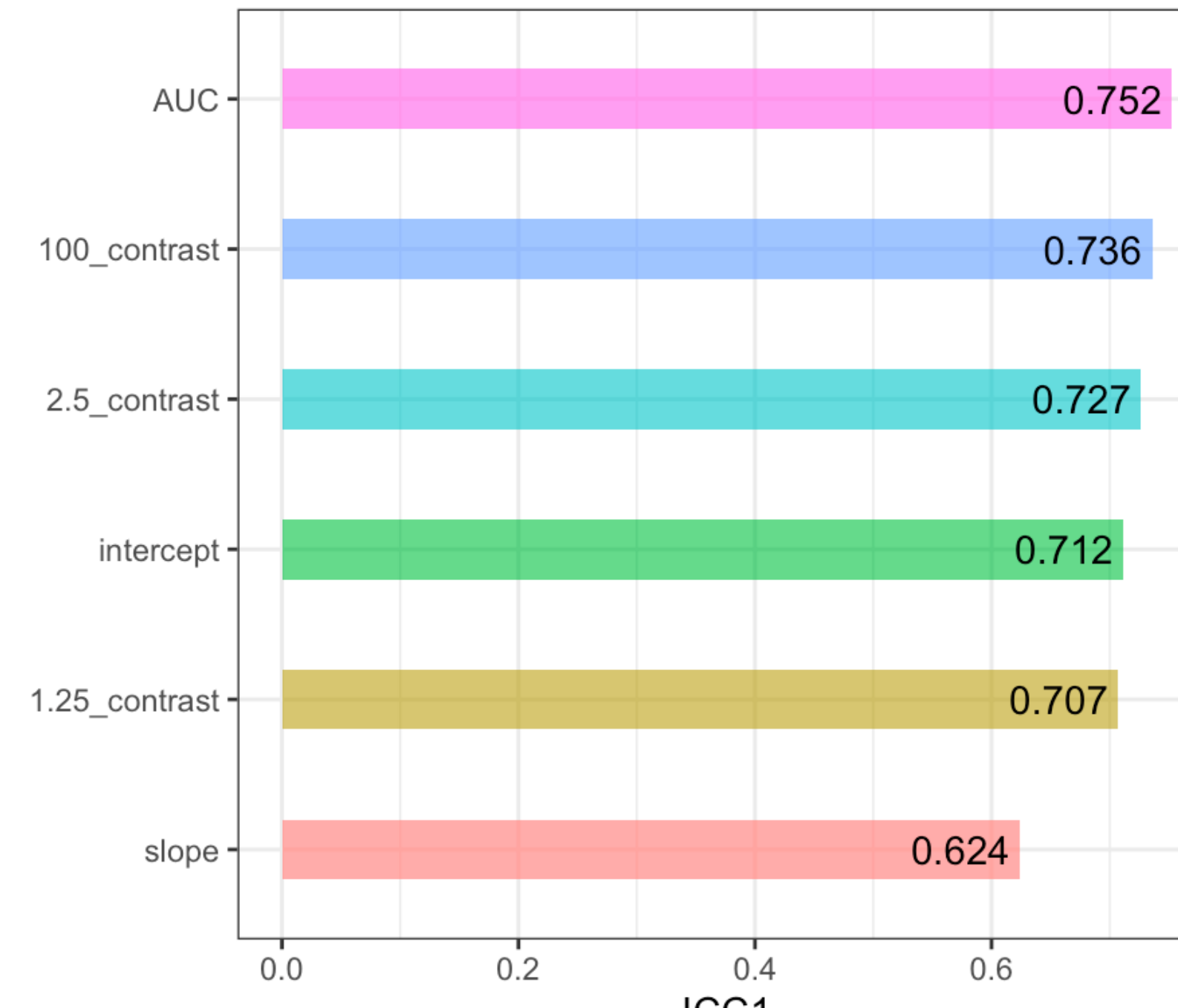


Figure 2: VA AUC outperforms other VA measures in intraclass correlation coefficient (ICC) analysis

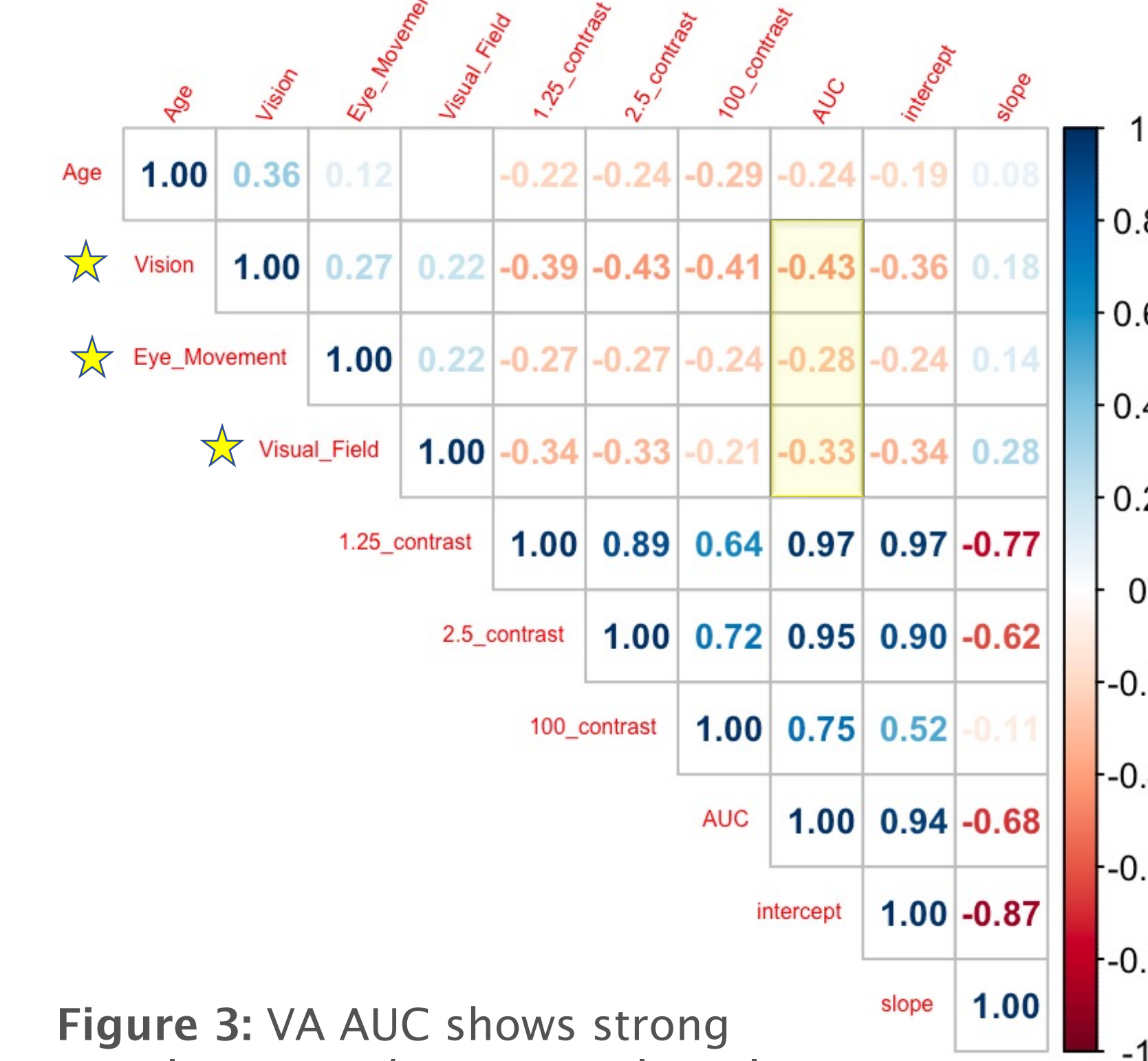


Figure 3: VA AUC shows strong correlations with existing clinical outcome measures

Conclusions

- VA AUC has higher ICC, greater dynamic range and stronger correlations with clinician-derived outcomes of visual disability than currently used 1.25% contrast
- OCT and VEP measurements have mostly non-overlapping effect on explanation of variance in the model
- Brain atrophy measures and OCT outcomes explain overlapping portion of variance
- There is additional variance in VA AUC not explained by the model, likely due to diseases anterior to retina or posterior visual pathway not reflected by brain atrophy

Results

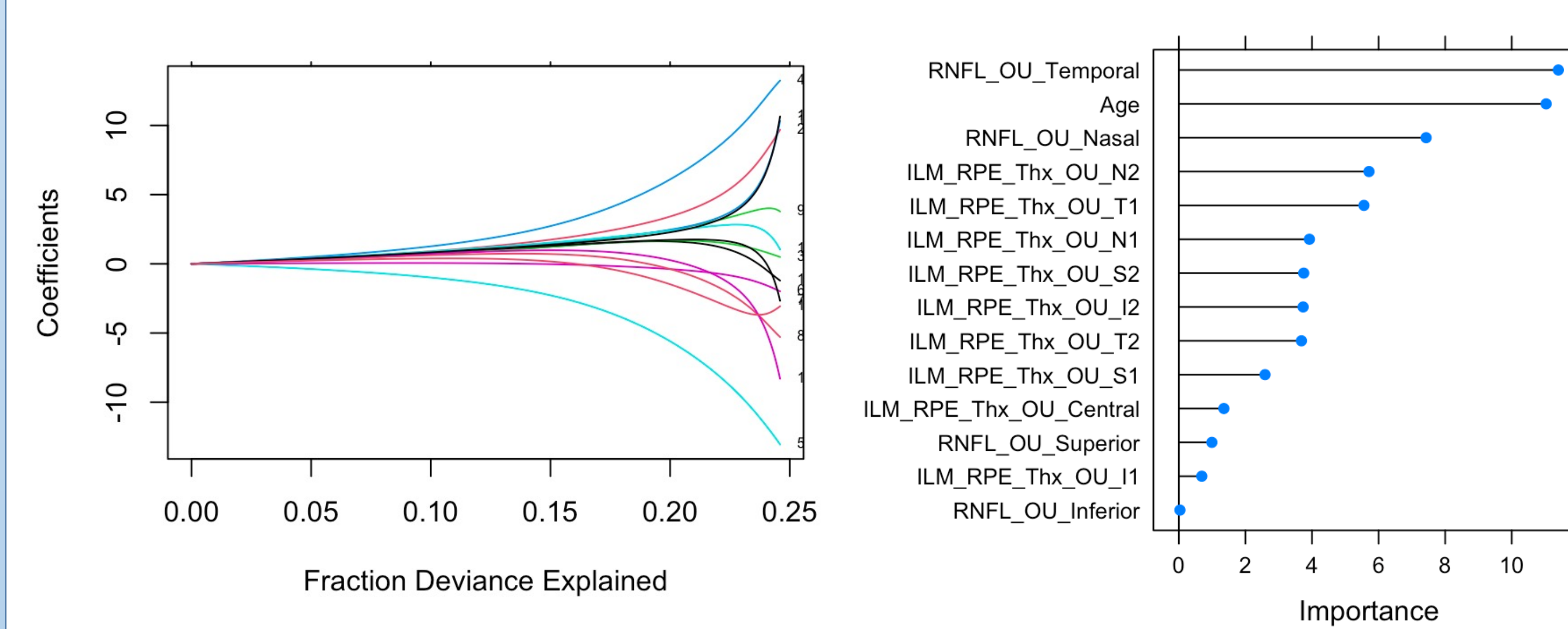


Figure 4: Model of VA AUC based on OCT parameters and age

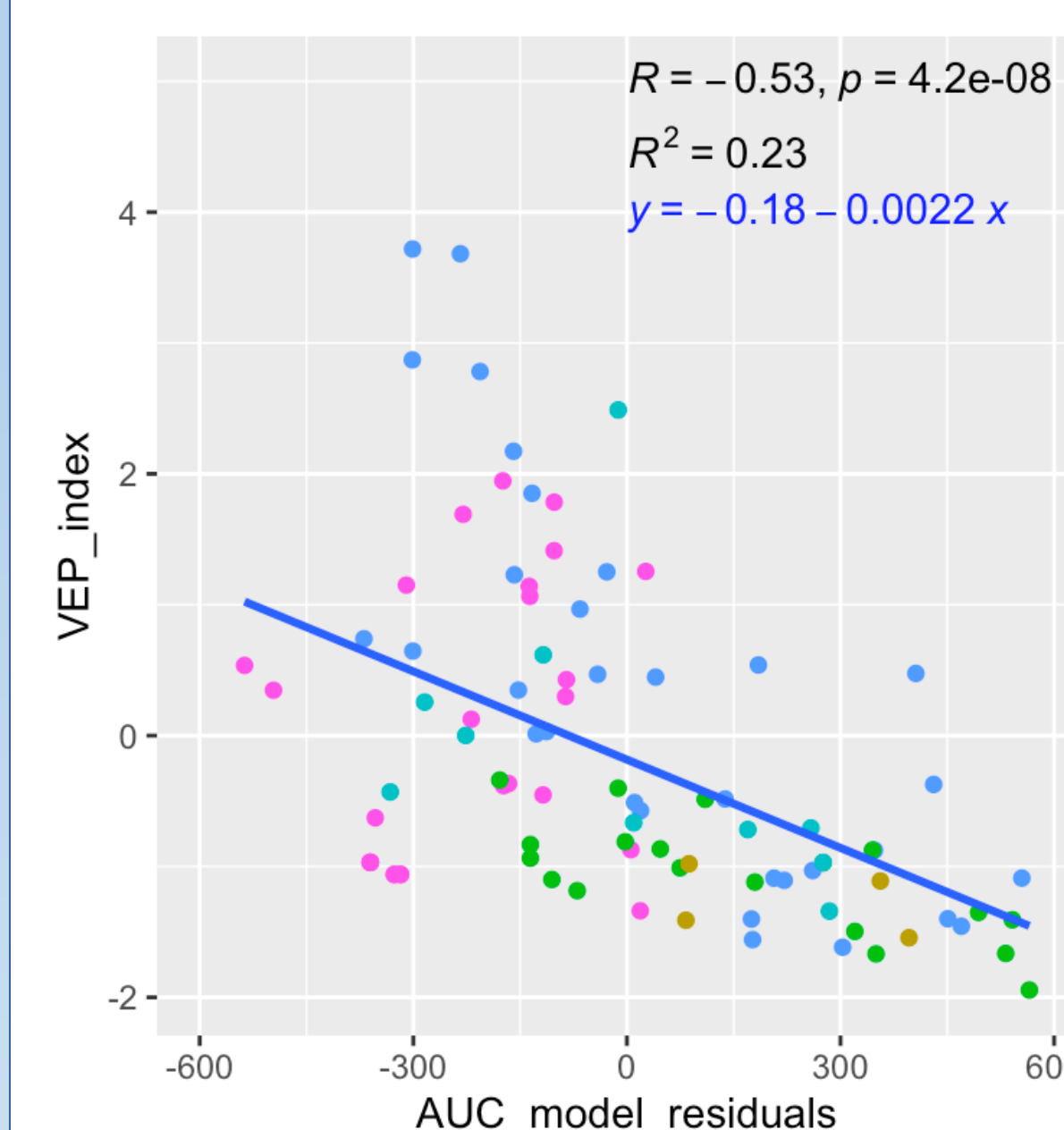


Figure 5: VEP explains portion of the residuals in OCT model

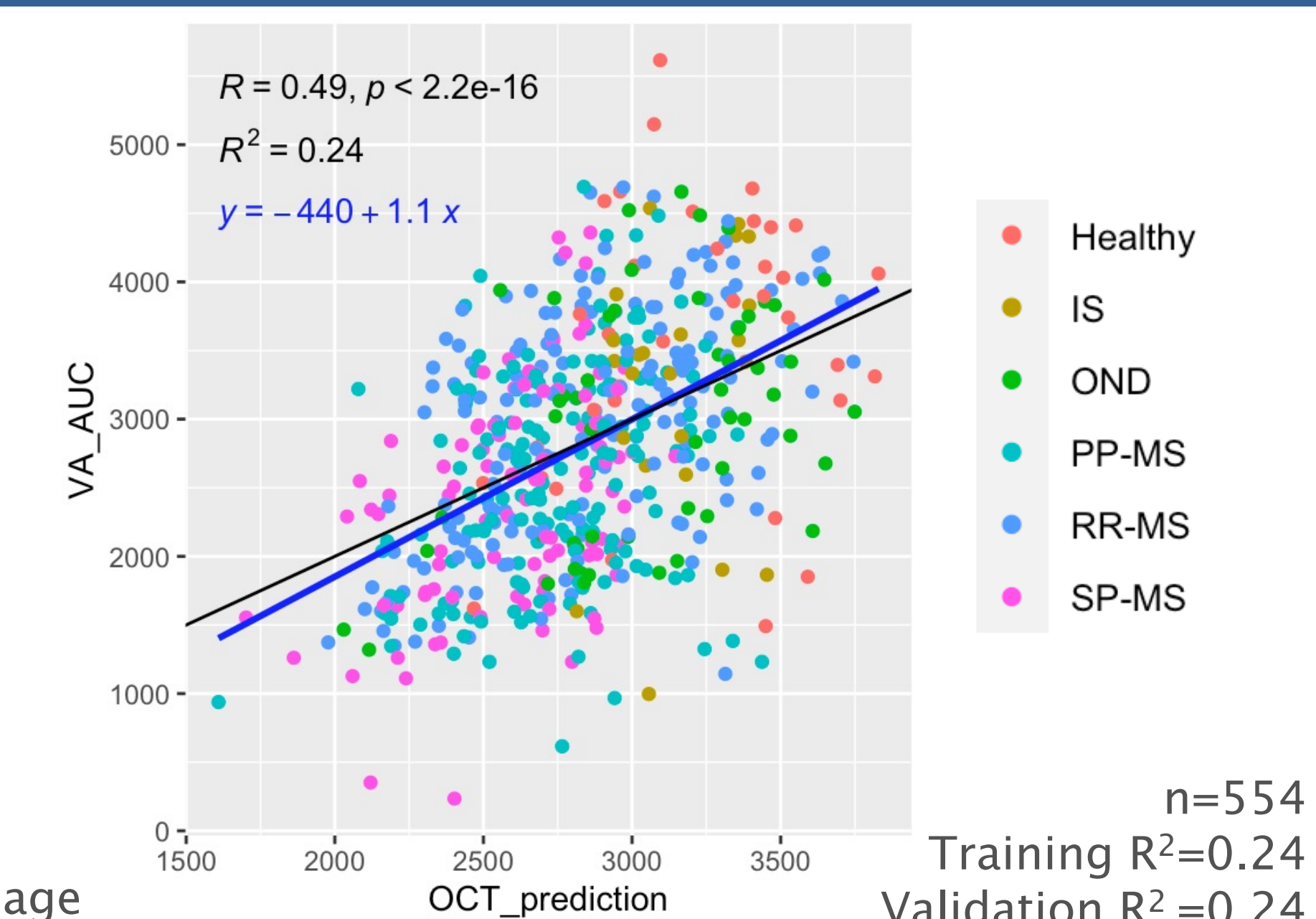
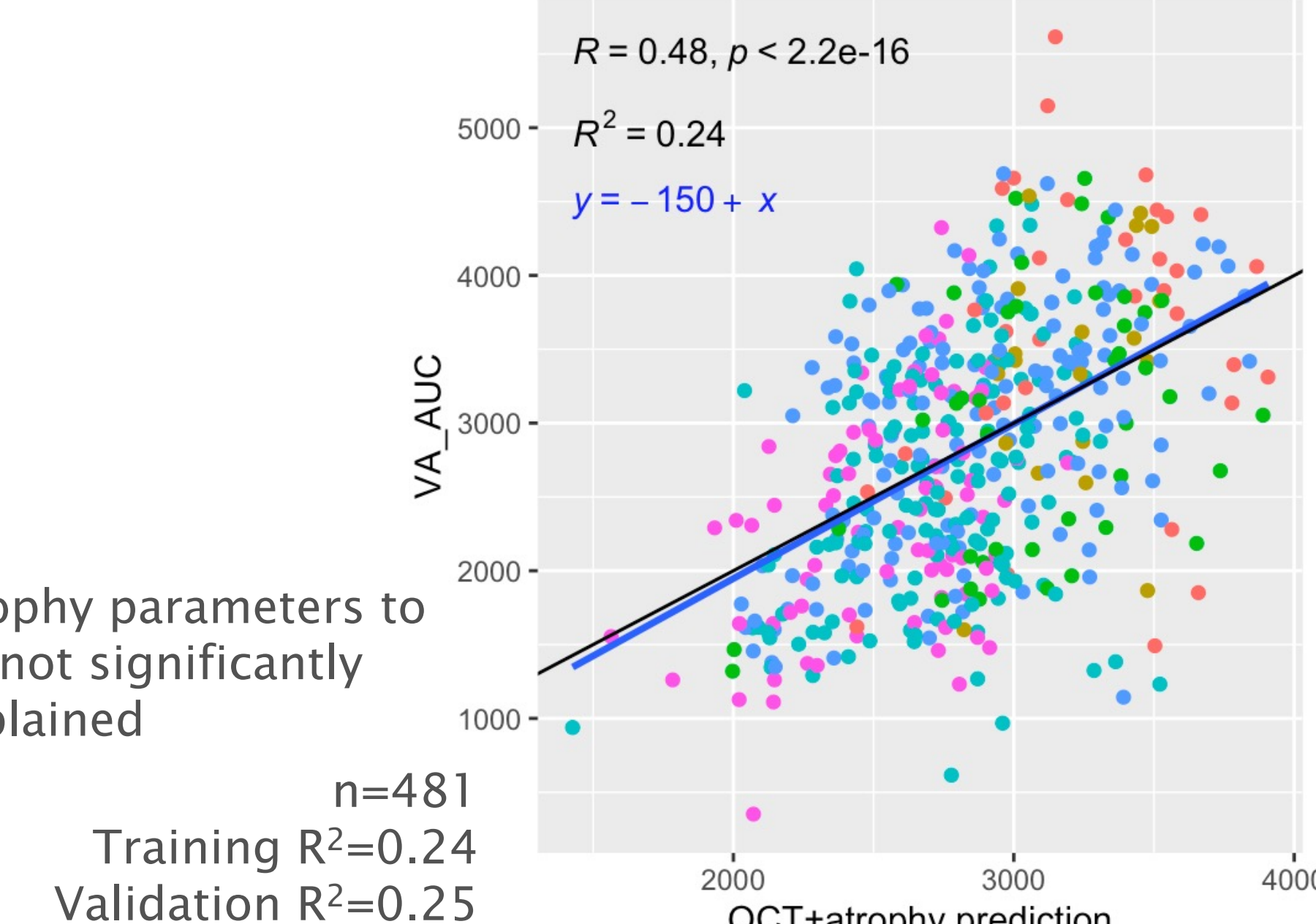


Figure 6: Adding atrophy parameters to the OCT model does not significantly increase variance explained



Future Directions

- Diffusion tensor imaging data focusing on visual pathway might explain more variance than BPFr
- Assess use of VA AUC in clinical practice for tracking disease progression
- Compare VA AUC from clinic and NDS developed smartphone app for patient monitoring

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