

Generalizability of Fluid Biomarkers

May 6, 2024

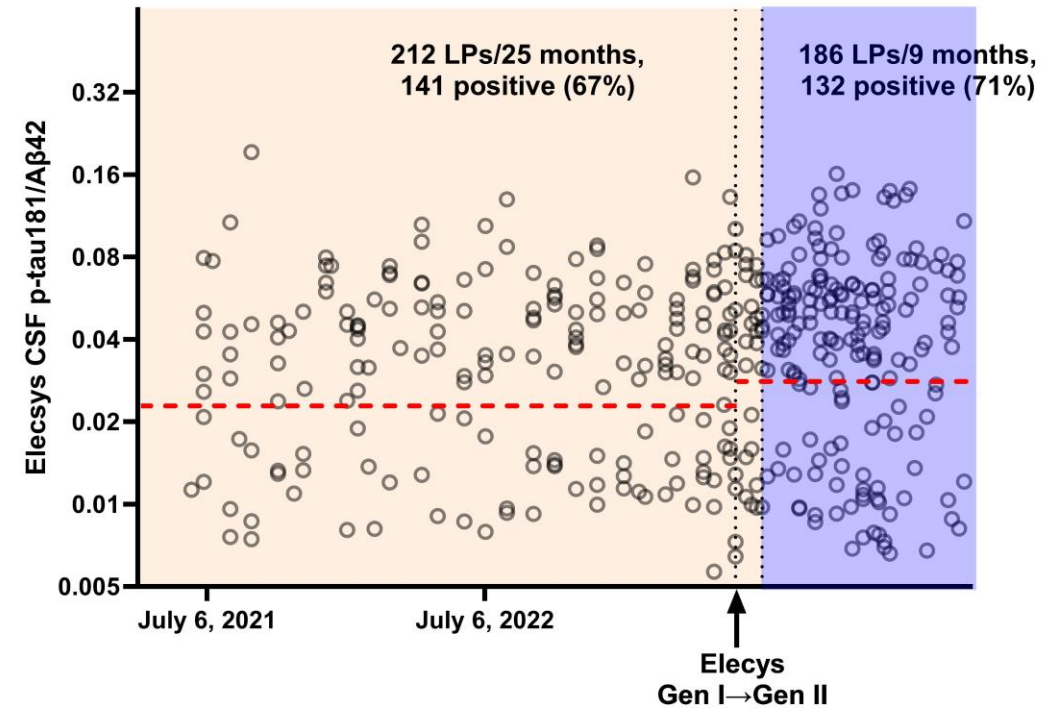
Disclosures: Suzanne Schindler, MD, PhD

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Why is generalizability of fluid biomarkers important?

- We are increasingly using fluid biomarkers in research, clinical trials and clinical diagnosis
- We are increasingly making treatment decisions based on the results of fluid biomarkers
- **The utility of a biomarker test depends on its ability to provide an accurate result for all individuals who undergo the test**

Clinical CSF testing at Washington University



Unpublished data

How do we assess the generalizability of a fluid biomarker test?

- Biomarkers are useful when levels strongly **reflect key biological and/or clinical factors** related to a disease of interest **Signal**
- Biomarker levels may additionally be affected by **factors not directly related to the disease of interest** that partially obscure the signal **Noise**
- Generalizability refers to **how consistently biomarkers reflect key biological and/or clinical factors across all individuals** **Signal:Noise**

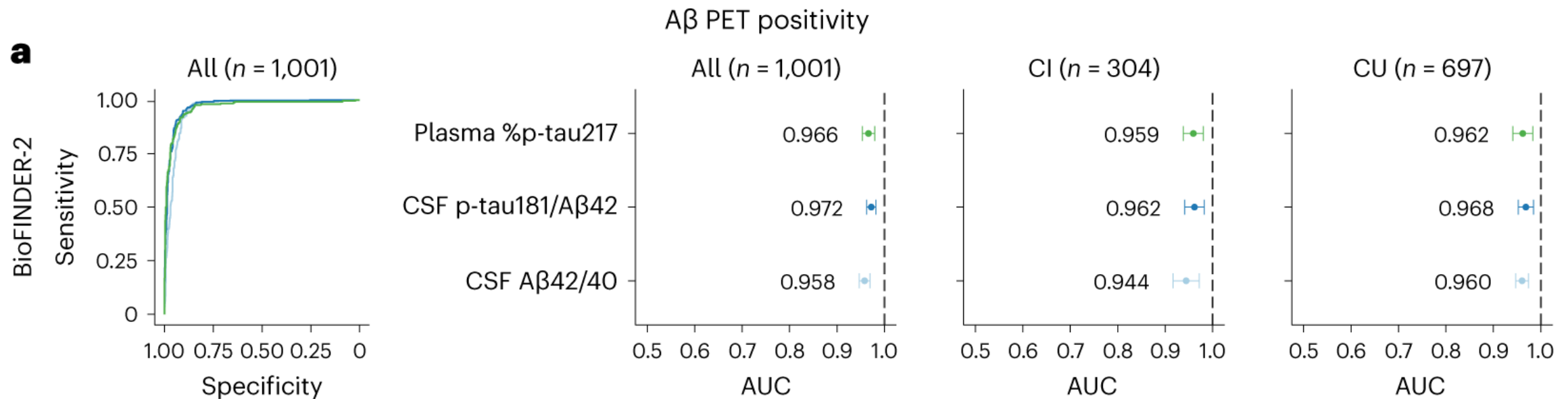
Low Signal:Noise High Signal:Noise



High generalizability requires a high signal:noise to clearly “see” the underlying biological/clinical condition, regardless of individual factors

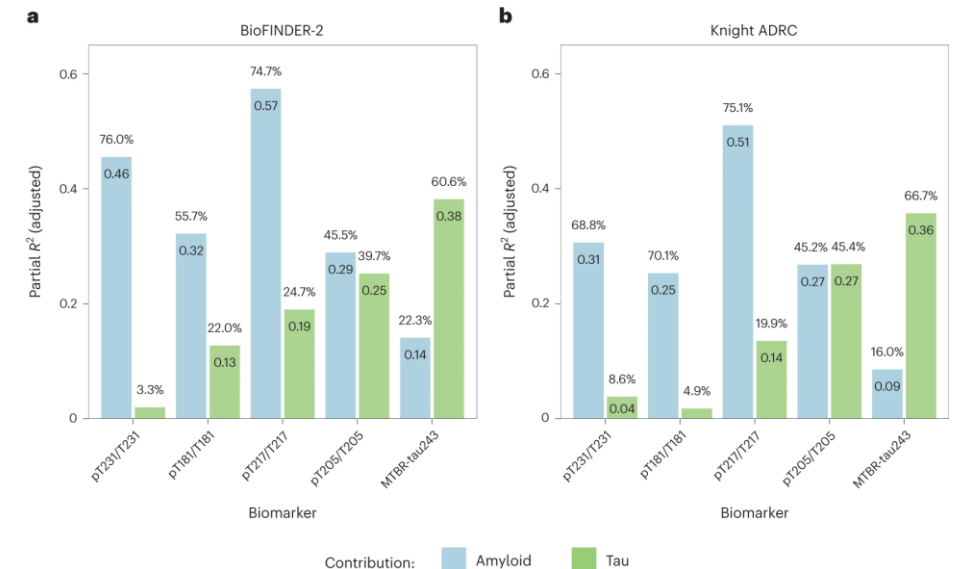
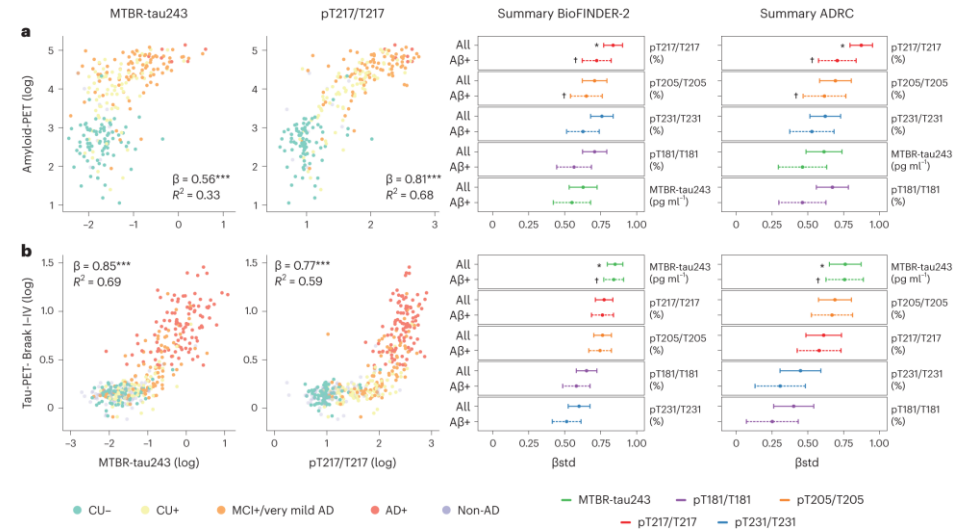
Evaluating the signal: classification accuracy

- An excellent “gold standard” for key biological and/or clinical factors is essential
- High associations between fluid biomarkers and the reference standard demonstrates a strong signal and high signal:noise

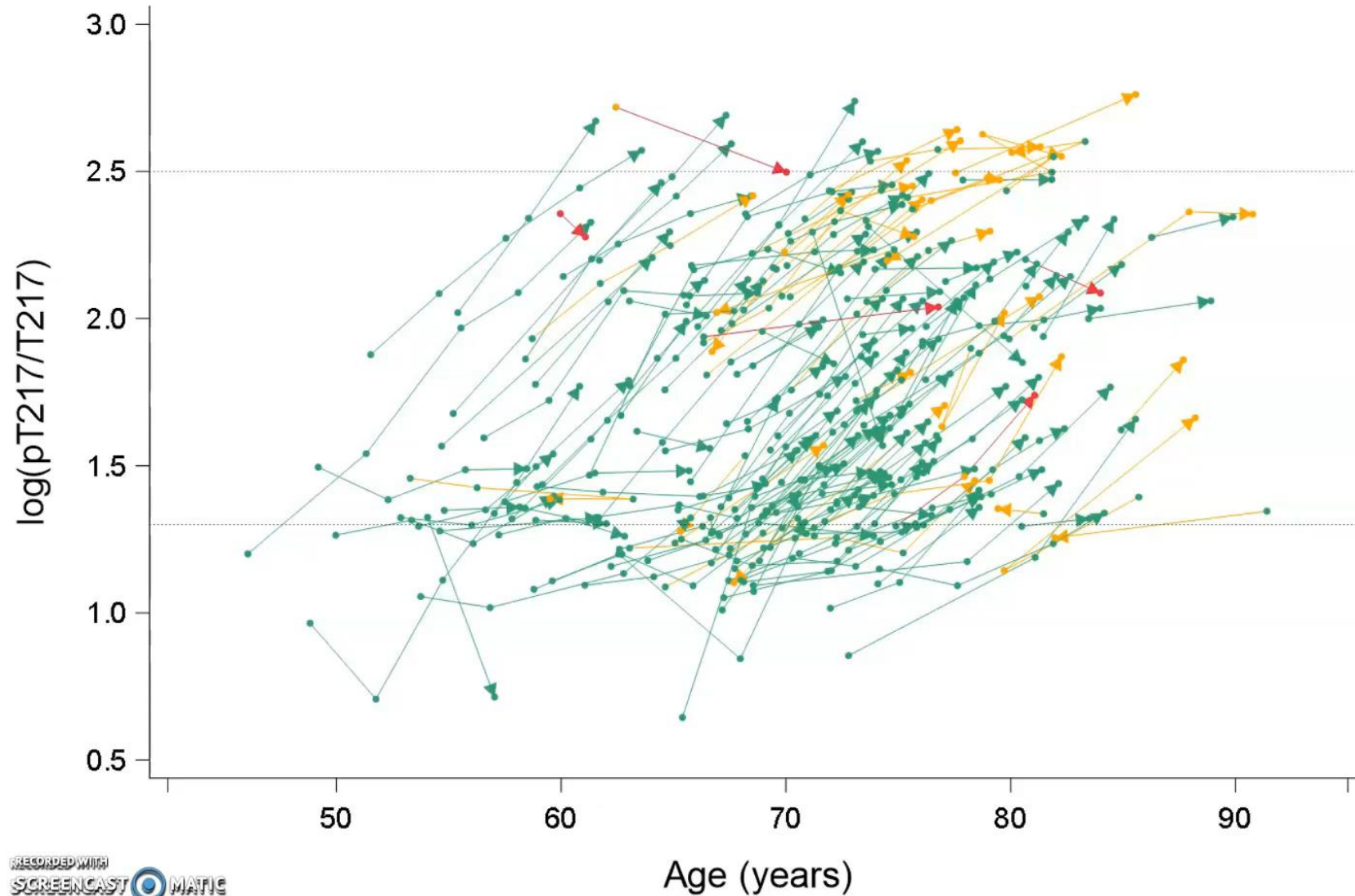


Evaluating the signal: continuous relationships

- Modeling can be performed to assess the strength of the association between continuous values for fluid biomarkers and reference standards
- Non-linear relationships are common, with stronger relationships during specific phases of disease
- Models can evaluate associations of fluid biomarkers with multiple pathologies simultaneously (e.g., amyloid and tau)



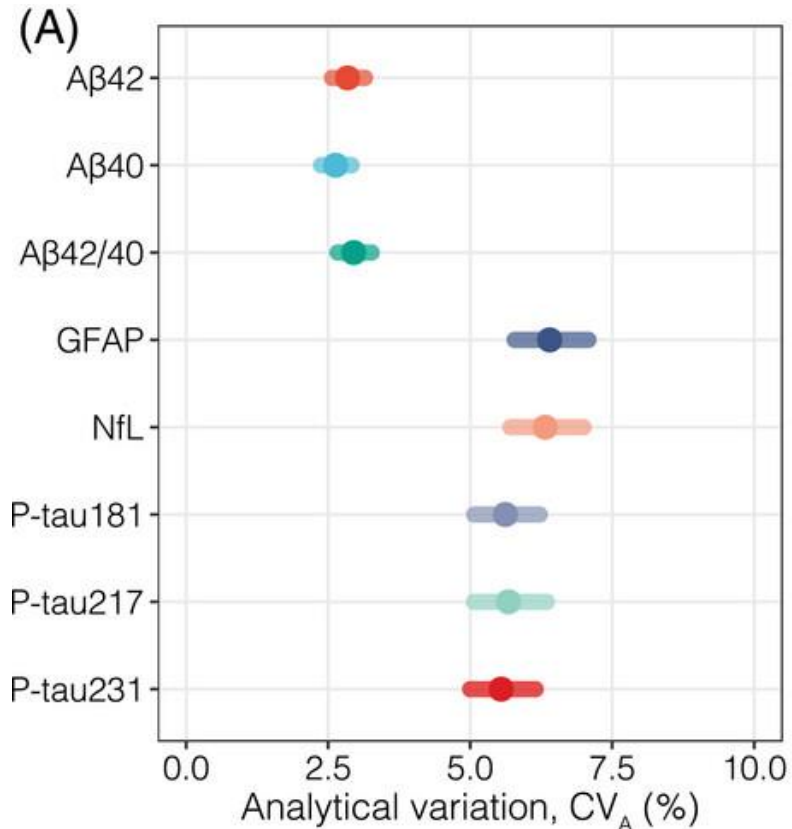
Evaluating the signal: longitudinal relationships



Evaluating the noise: cross-sectional and longitudinal variance

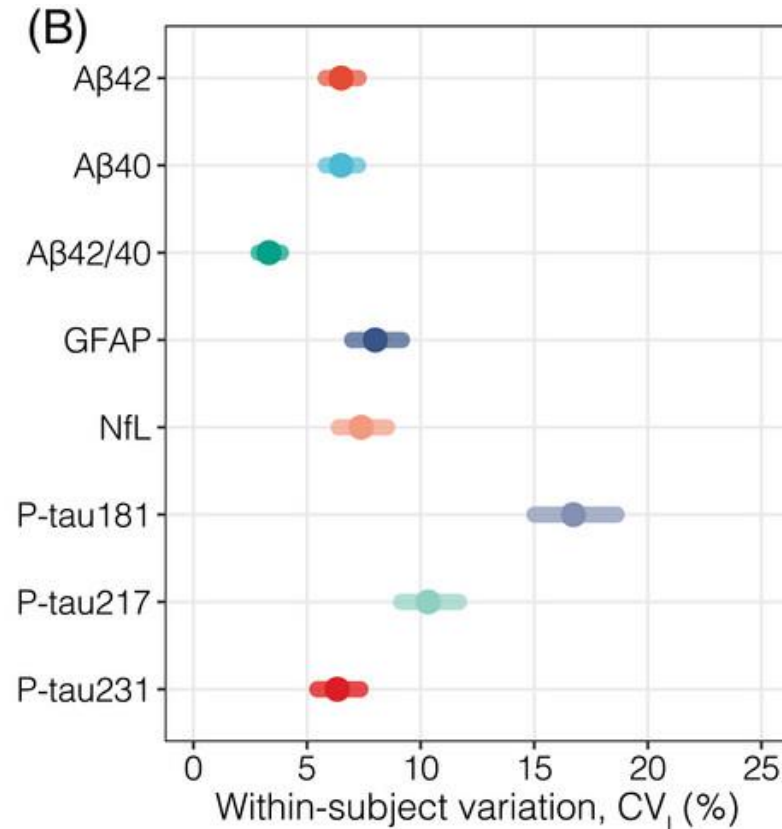
Noise

Analytical variation



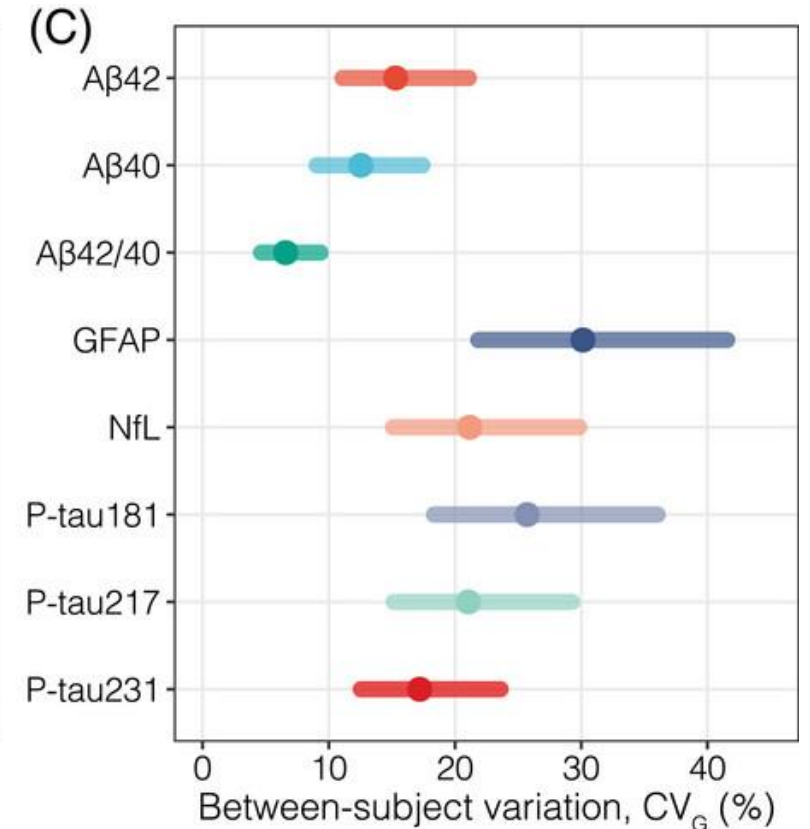
Noise

Intra-individual variation



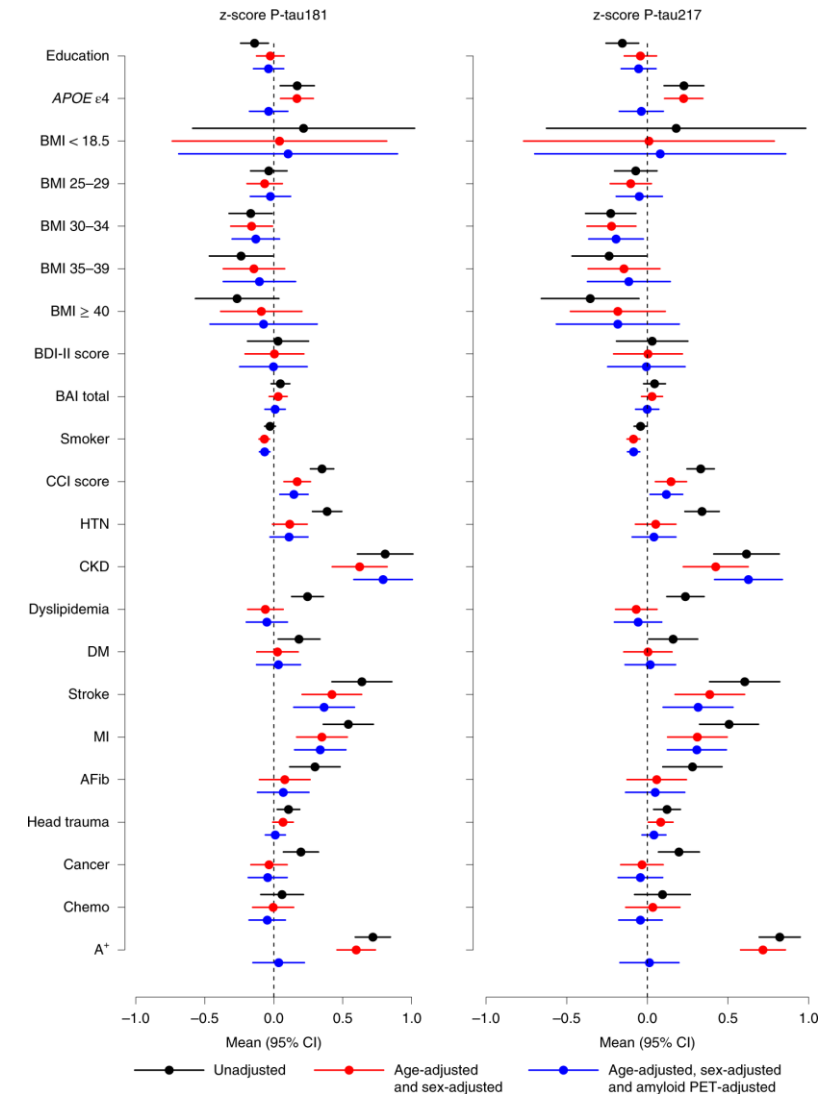
Signal + Noise?

Related + un-related to disease



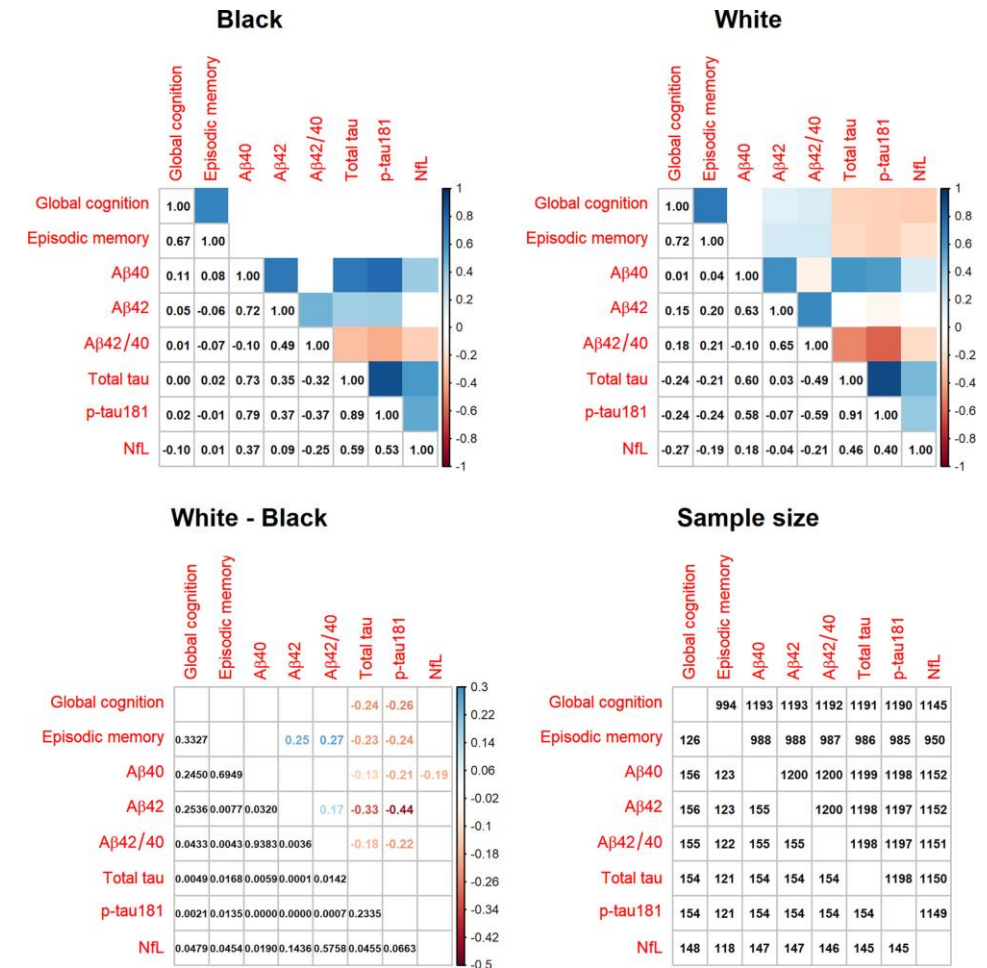
Evaluating the noise: continuous relationships

- Levels of some fluid biomarkers are associated with factors not directly related to the disease of interest (e.g., age, sex, race, ethnicity, comorbidities, medications)
- Especially when lower performing biomarker tests are used in a diverse clinical populations (e.g., Bouteloup *et al.*, *Neurology* 2024), the signal:noise may be extremely low, such that the biomarker test result has little relationship to the disease and depends largely on non-disease related factors



Evaluating the noise: differences in continuous relationships

- The correlations between CSF biomarkers and cognitive measures vary between Black and White individuals
- Rates of amyloid positivity may vary by race/ethnicity, even after accounting for many other variables
- Lower rates of amyloid positivity may exacerbate under-representation of Black and Hispanic individuals in clinical trials (Molina-Henry, *et al.*, *Alzheimer's and Dementia* 2024)



Evaluating the noise: classification accuracy

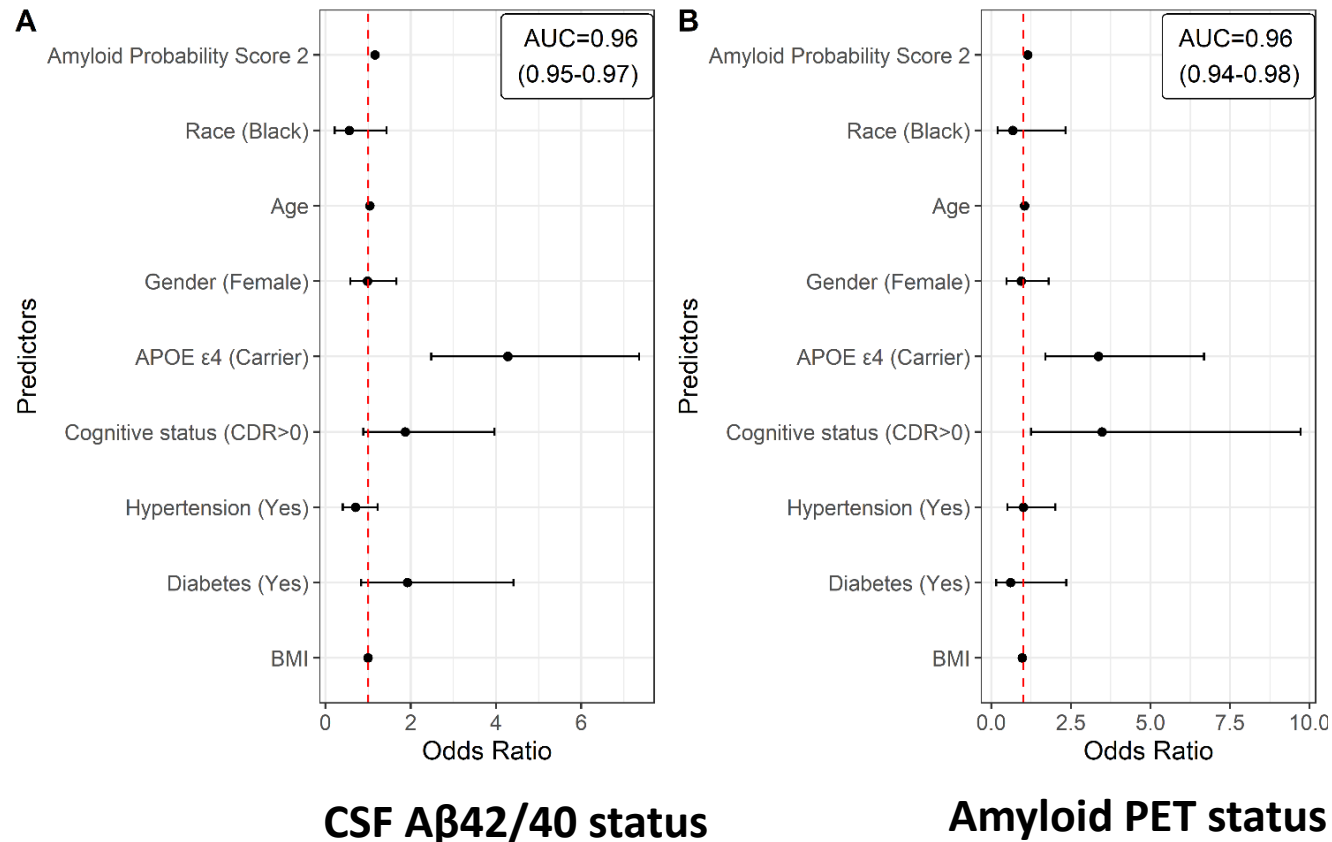
- Logistic regression models of amyloid status
 - Covariate-adjusted biomarker effects reflect the signal
 - Covariate effects may (or may not) reflect noise
 - The AUC reflects the signal:noise
- For biomarker tests with high signal:noise (AUC)
 - Biomarkers have stronger associations with amyloid status
 - Covariates have weaker associations with amyloid status
- For biomarker tests with low signal:noise (AUC)
 - Biomarkers have weaker associations with amyloid status
 - Covariates have stronger associations with amyloid status
- **High accuracy tests are needed for generalizability**

Classification of amyloid status

Plasma A β 42/A β 40, AUC 0.90 (0.85-0.96)			
Parameter	Estimate	SE	p =
Intercept	13.0	4.7	0.005
Plasma A β 42/A β 40 (pg/ml)	-220	46	<0.0001
Race (African American)	0.058	0.274	N.S.
Sex (female)	0.843	0.568	N.S.
Age (years)	0.109	0.04	0.007
APOE ϵ 4 status (carrier)	0.865	0.269	0.001
Cognitive status (CDR>0)	1.11	0.41	0.007
Plasma p-tau181, AUC 0.85 (0.79-0.92)			
Parameter	Estimate	SE	p =
Intercept	-8.69	2.71	0.001
Ln (plasma p-tau181)	1.53	0.57	0.007
Race (African American)	-0.59	0.22	0.007
Sex (female)	-0.21	0.44	N.S.
Age (years)	0.072	0.035	0.04
APOE ϵ 4 status (carrier)	0.87	0.23	0.0002
Cognitive status (CDR>0)	1.02	0.39	0.009
Plasma NfL, AUC 0.81 (0.74-0.89)			
Parameter	Estimate	SE	p =
Intercept	-6.20	2.41	0.01
Ln (plasma NfL)	-0.097	0.476	N.S.
Race (African American)	-0.65	0.22	0.003
Sex (female)	-0.50	0.42	N.S.
Age (years)	0.109	0.040	0.007
APOE ϵ 4 status (carrier)	0.89	0.23	<0.0001
Cognitive status (CDR>0)	1.27	0.39	0.001

A high accuracy blood test classifies amyloid status consistently

- Individuals in the Knight ADRC cohort (75 Black individuals, 687 White individuals)
- Race, sex, age, hypertension, diabetes, and BMI did not significantly affect classification of amyloid status by APS2 (plasma A β 42/40 + %p-tau217)



Needed: more representative cohorts

- Research cohorts traditionally have been relatively homogenous, with potentially lower biomarker variance due to non-disease related factors
- Ideally, cohorts used for biomarker validation would resemble the intended use population
 - Community-based for screening tests
 - Clinic-based for diagnostic tests
- However, biomarker validation requires high quality reference standards (CSF and PET), and it may be difficult to obtain CSF and PET in a truly representative cohorts
- Smaller, targeted, and carefully designed studies to evaluate the effects of certain conditions may be more efficient than very large studies of low frequency conditions

Conclusions

- We need fluid biomarkers that accurately and consistently reflect disease processes in all individuals
- High generalizability requires a high signal:noise to clearly “see” the underlying biological/clinical condition
- A variety of studies can be used to assess the signal:noise of different fluid biomarker tests
- We need more representative cohorts with reference standards to validate fluid biomarkers
- Smaller, carefully designed cohorts may also be helpful in answering specific questions

Acknowledgements

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Knight ADRC Fluid Biomarker Core

