

Joint meta-analysis of two diagnostic tests using bivariate copulas to model within-study dependencies in Health Technology Assessment (HTA) of novel biomarkers in Alzheimer's disease dementia

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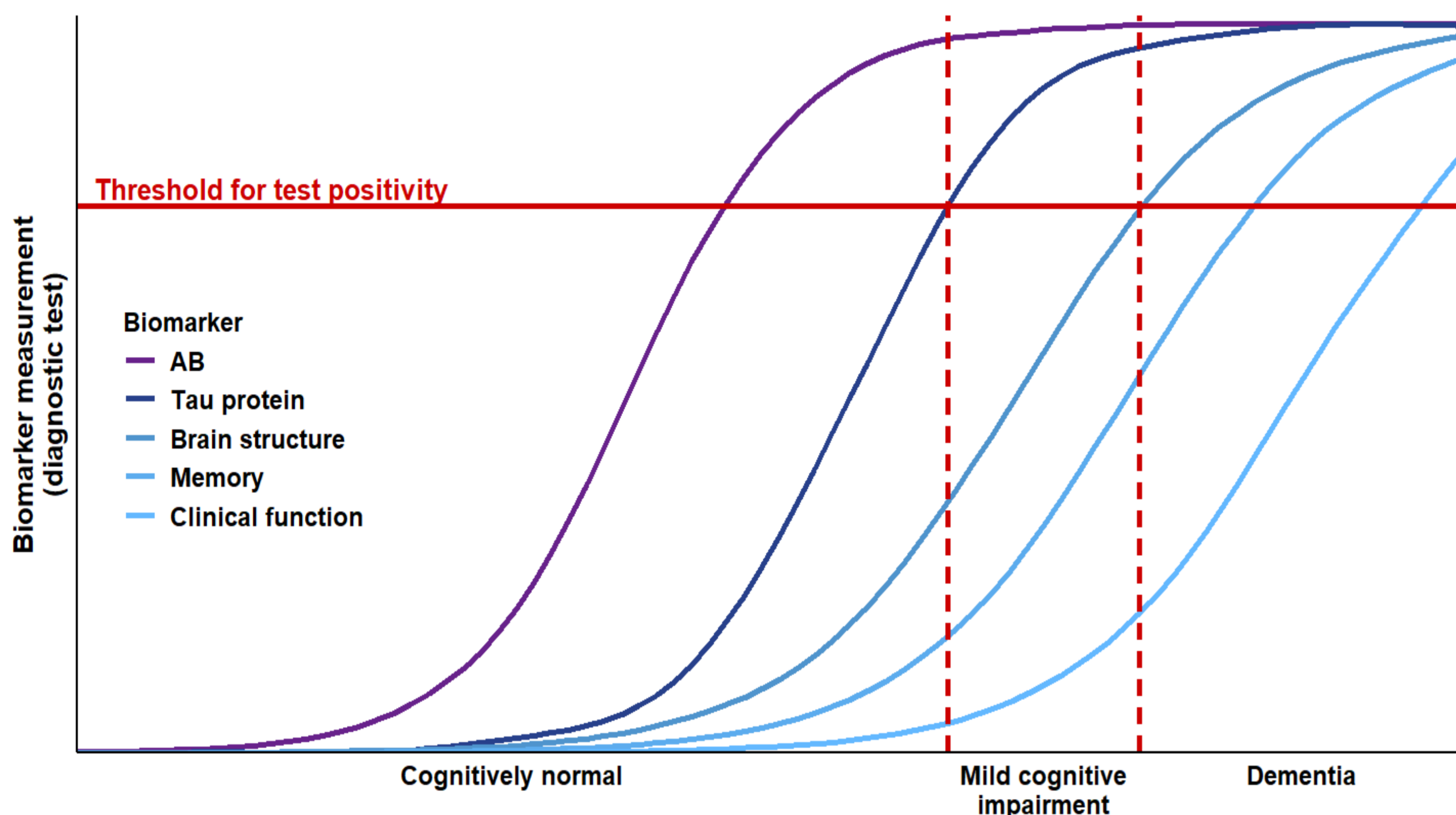
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Background

- Health Technology Assessments (HTAs) considered by the National Institute for Health and Care Excellence (NICE) Diagnostics Assessment Programme often concern the **comparative or combined accuracy of two or more diagnostic tests**. [1] Their estimation requires modelling techniques that account for **within-study correlations** between the tests.
- We describe a novel application of a **meta-analytic model with copulas** [2] to capture within-study dependencies between two tests assessed in the same patients, using **study- or individual-level data** where available.
- The methodology is applied to a motivating example assessing the accuracy of emerging biomarkers in **Alzheimer's disease dementia** (Figure 1).

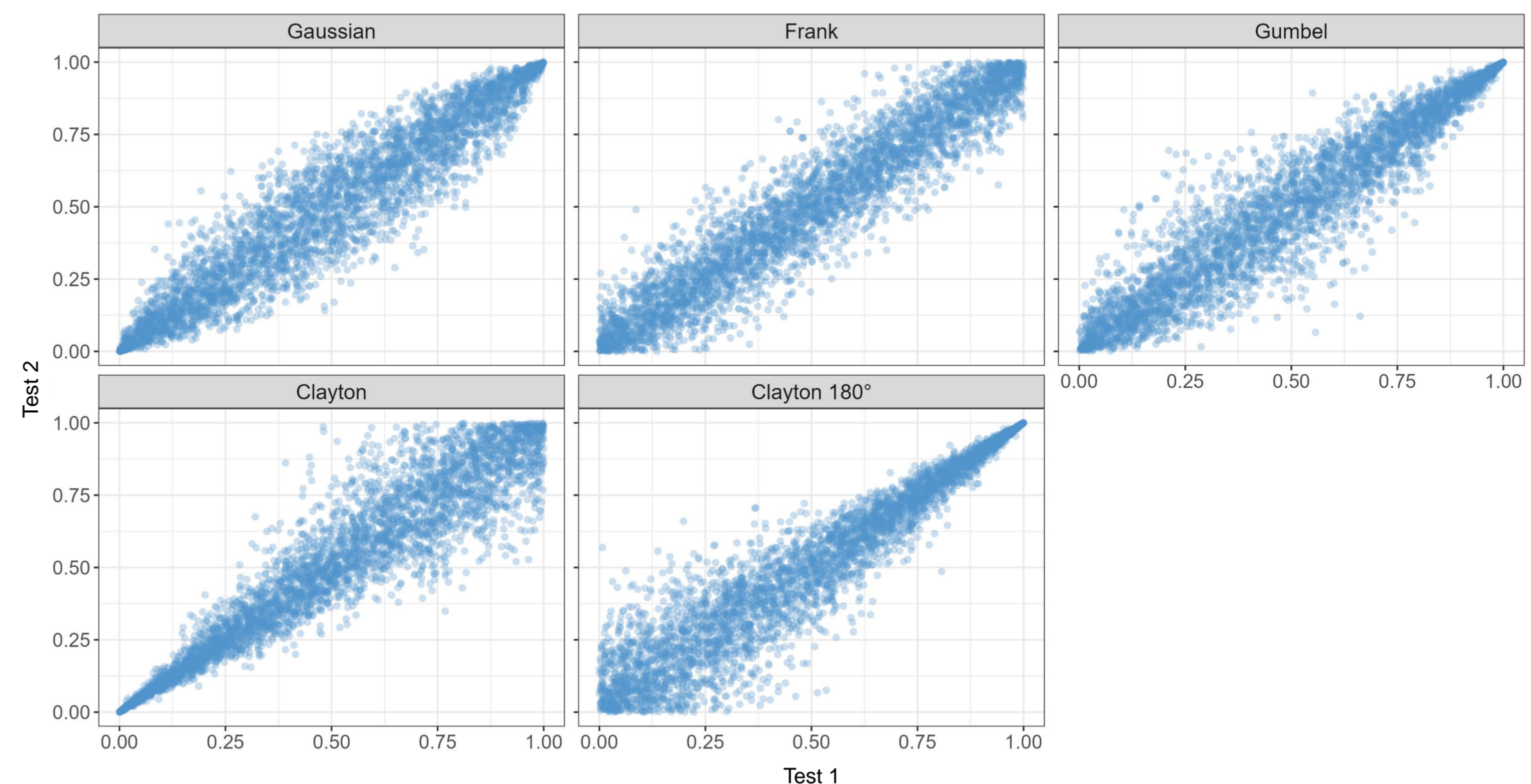
Figure 1: Theoretical progression of different biomarkers over time, adapted from a schematic diagram by Jack et al (2010). [3]



Aim: Develop novel meta-analysis models for jointly synthesising diagnostic accuracy data on two tests for Alzheimer's disease dementia

Methods

Figure 2: Simulated samples from the different types of copulas.



- Novel **Bayesian meta-analysis models** for evaluating the accuracy of two diagnostic tests in the same patients were developed using a motivating example in Alzheimer's disease dementia.
- Bivariate copulas** were used to **flexibly capture within-study dependencies** between the two tests, relaxing the need for individual participant data from all studies.
- Five bivariate copula models** capturing different dependence structures were fit to the data: Gaussian, Frank, Gumbel, Clayton and Clayton rotated 180° (Figure 2).
- The models were compared to the **currently recommended meta-regression approach** [4] for modelling two cerebrospinal fluid biomarker (CSF) tests for diagnosing Alzheimer's disease dementia: amyloid- β 42 ($A\beta_{42}$) and total tau (t-tau).
- Model fit was assessed using the widely applicable information criterion (WAIC). [5]

Results

- CSF $A\beta_{42}$ and t-tau demonstrated **sensitivities** of 80.9% (95% credible interval: 73.4%, 87.5%) and 76.4% (69.4%, 83.1%), respectively.
- Summary **specificity** was 70.3% (61.3%, 78.4%) and 72.5% (63.7%, 81.3%), respectively.
- The bivariate copula models resulted in a **better fit** compared to the meta-regression model, and **increased precision** in estimates of sensitivity and specificity by as much as a **15% reduction in the width of the 95% credible intervals** (Figure 3).

Conclusions

- The bivariate copula framework **supports HTA**, enabling test comparisons while accounting for **complex dependence structures** arising between tests.
- Increased precision in sensitivity and specificity estimates aids the evaluation of **clinical and cost-effectiveness of diagnostic tests**, enabling more appropriate decisions regarding the most **efficient use of health resources**.
- This novel methodological development is applicable to a **broad range of disease areas**.

Figure 3: Posterior medians (solid dots) and 95% CrIs (solid bars) of test accuracy parameters.

