

# Model-based subtypes of disease progression in Parkinson's

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**A data-driven take on understanding heterogeneity in Parkinson's disease.**  
**Subtype and Stage Inference (SuStaln):**  
**Computational model of disease progression + Clustering**

## Parkinson's disease (PD)

- 2<sup>nd</sup> most common neurodegenerative disease
- Highly heterogeneous

- ∃ no validated biomarkers of progression for PD
- Needed for clinical trials

**We estimate a data-driven signature of PD progression subtypes as sequences of measurable abnormality in the Parkinson's Progression Markers Initiative (PPMI)**

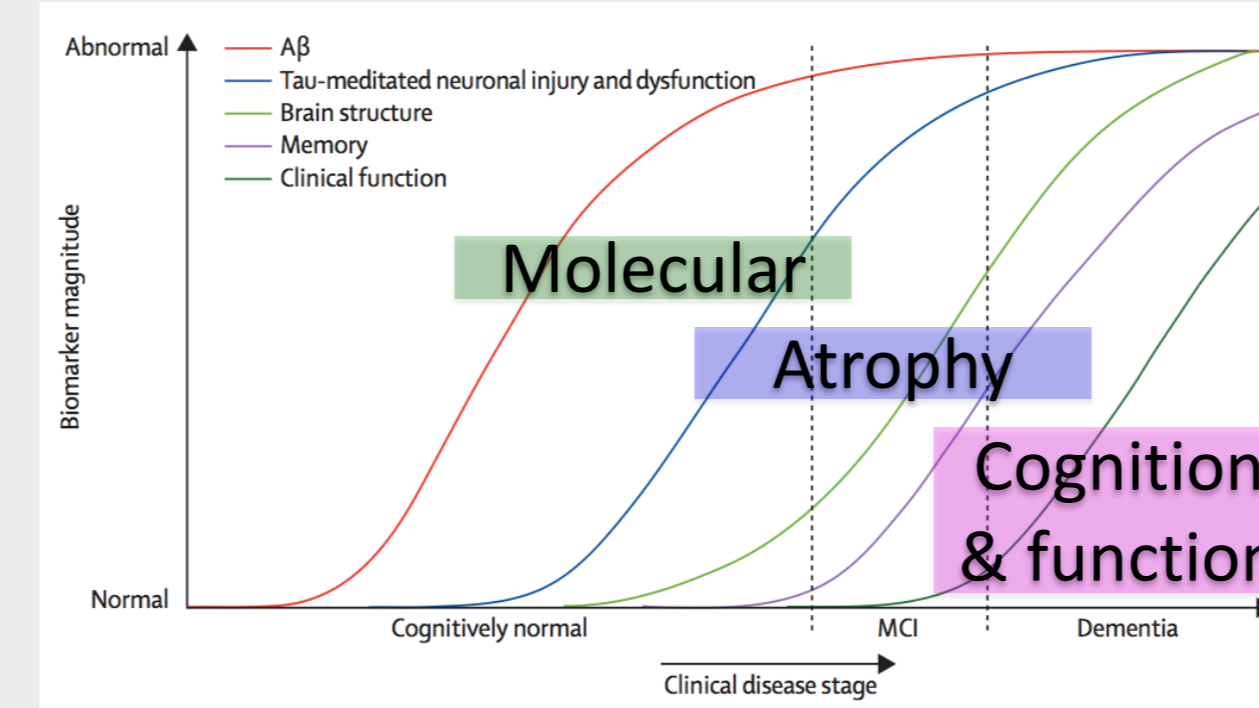
## Study overview: SuStaln + KDE-EBM

- Event-based model (EBM) [1,2] estimates a probabilistic sequence of cumulative abnormality in a set of N features using mixture modelling to quantify abnormality
- KDE-EBM [3], Kernel Density Estimation mixture model
- **Methodological novelty here: variable bandwidth KDE**
- SuStaln [4]: split-and-fit KDE-EBM clusters
- Patient subtype and stage inferred from best fitting (max. likelihood) combination of abnormal features
  - "Stage M" ≅ M out of N markers are abnormal

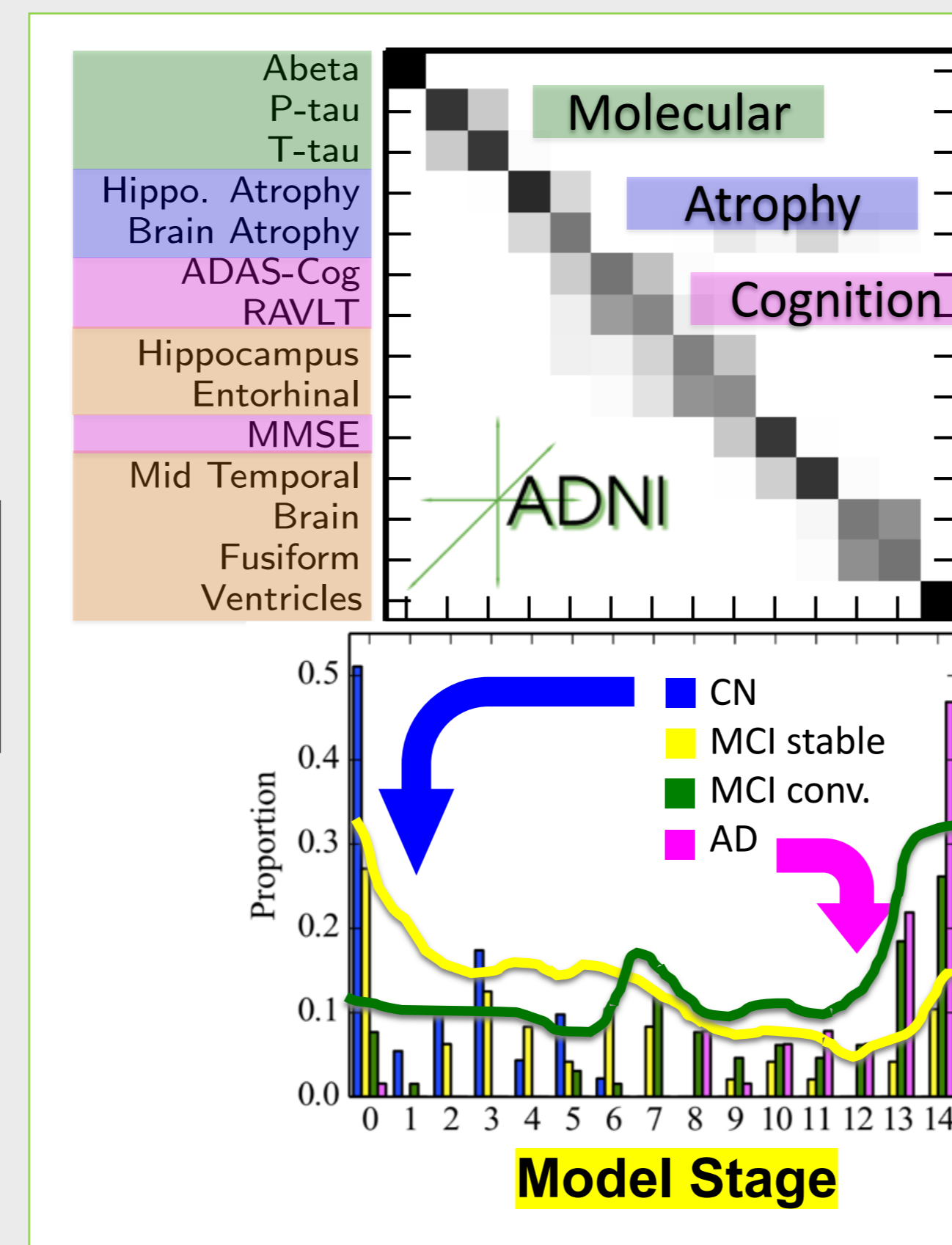
## Methods and Data

**Event Based Model (EBM) [1,2]** estimates a probabilistic sequence of biomarker abnormality => fine-grained staging tool

inspired by hypothetical model: (Alzheimer's) [5]

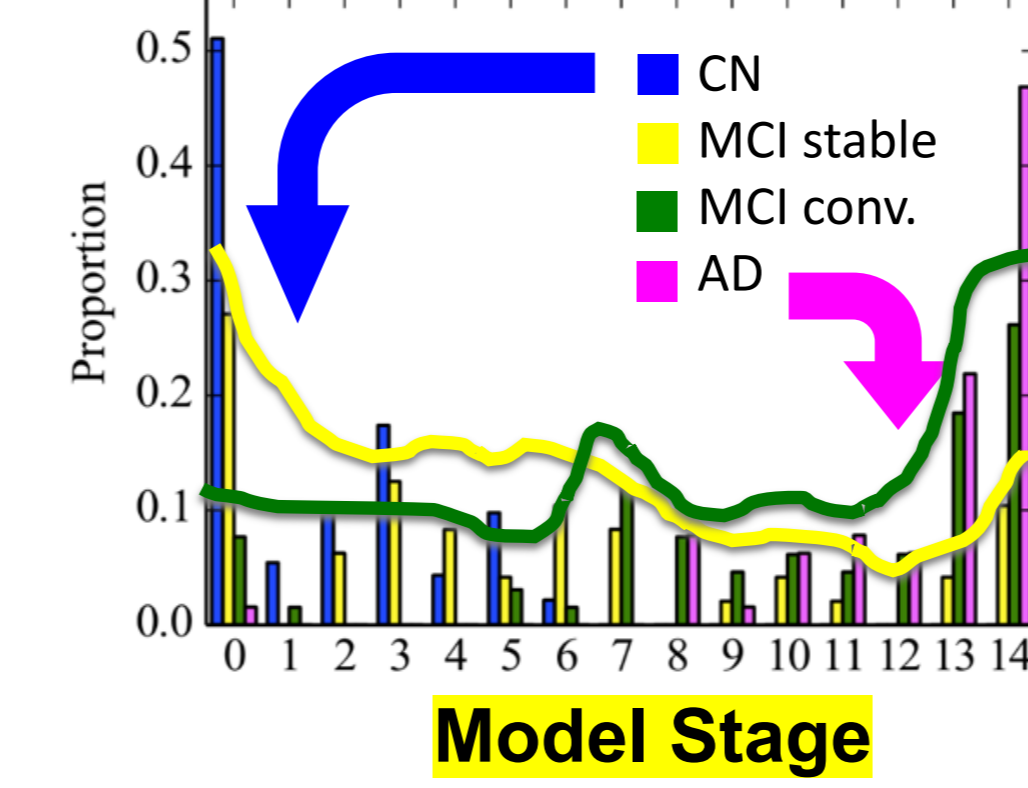


developed a data-driven model: [1,2]

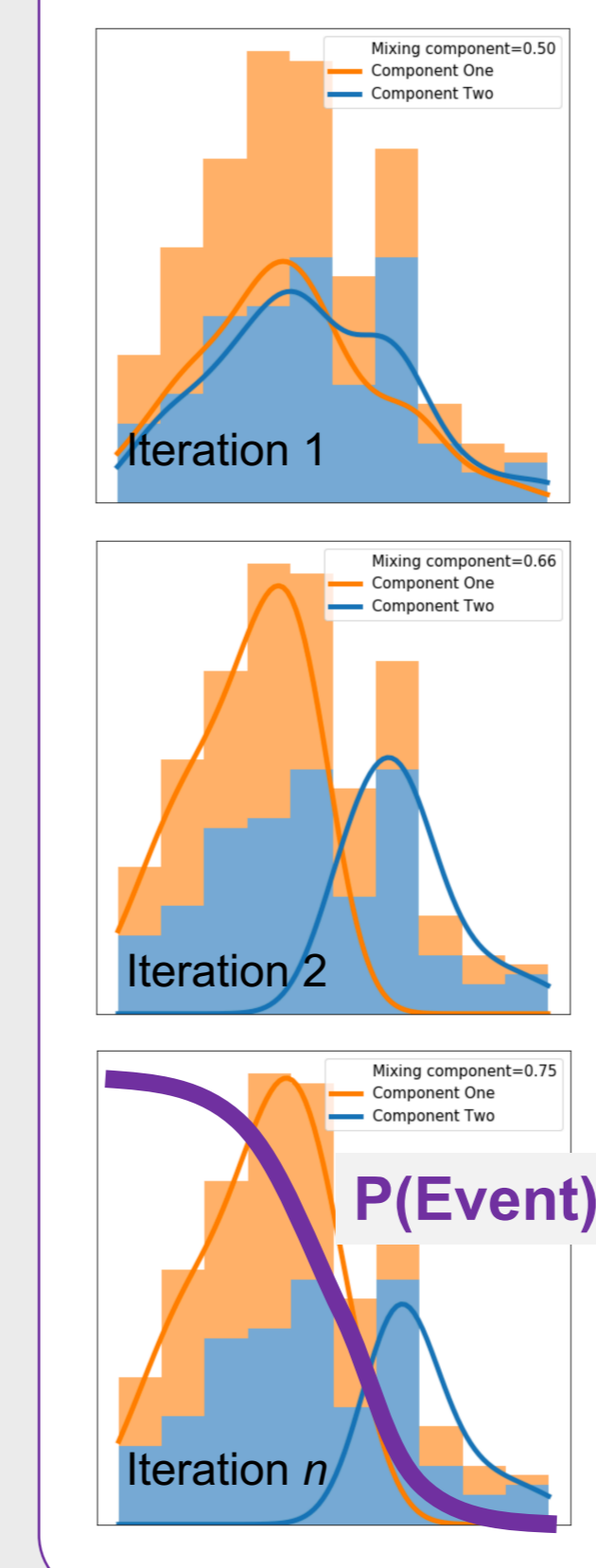


Key: Grayscale intensity (per row) represents the likelihood of the biomarker event occurring in that sequence position

which provides a staging tool: [2]



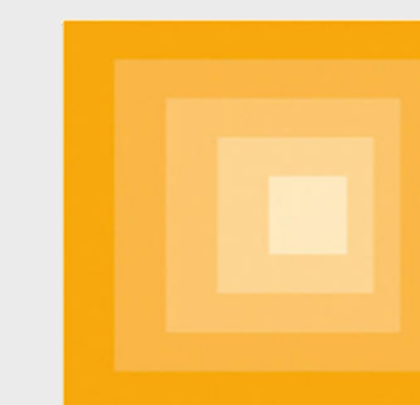
## KDE Mixture Modelling [3]



## Data

- 9 "Imaging plus X" [6] features: clinical, cognitive, DaTscan SPECT striatal binding ratio
- Inclusion criteria: complete data

Group	N	Disease duration	Age (years)
HC	175	n/a	61 ± 11
de novo PD	405	0.6 ± 0.5 years	62 ± 10
SWEDD	56	0.6 ± 0.7 years	62 ± 10



PARKINSON'S PROGRESSION MARKERS INITIATIVE

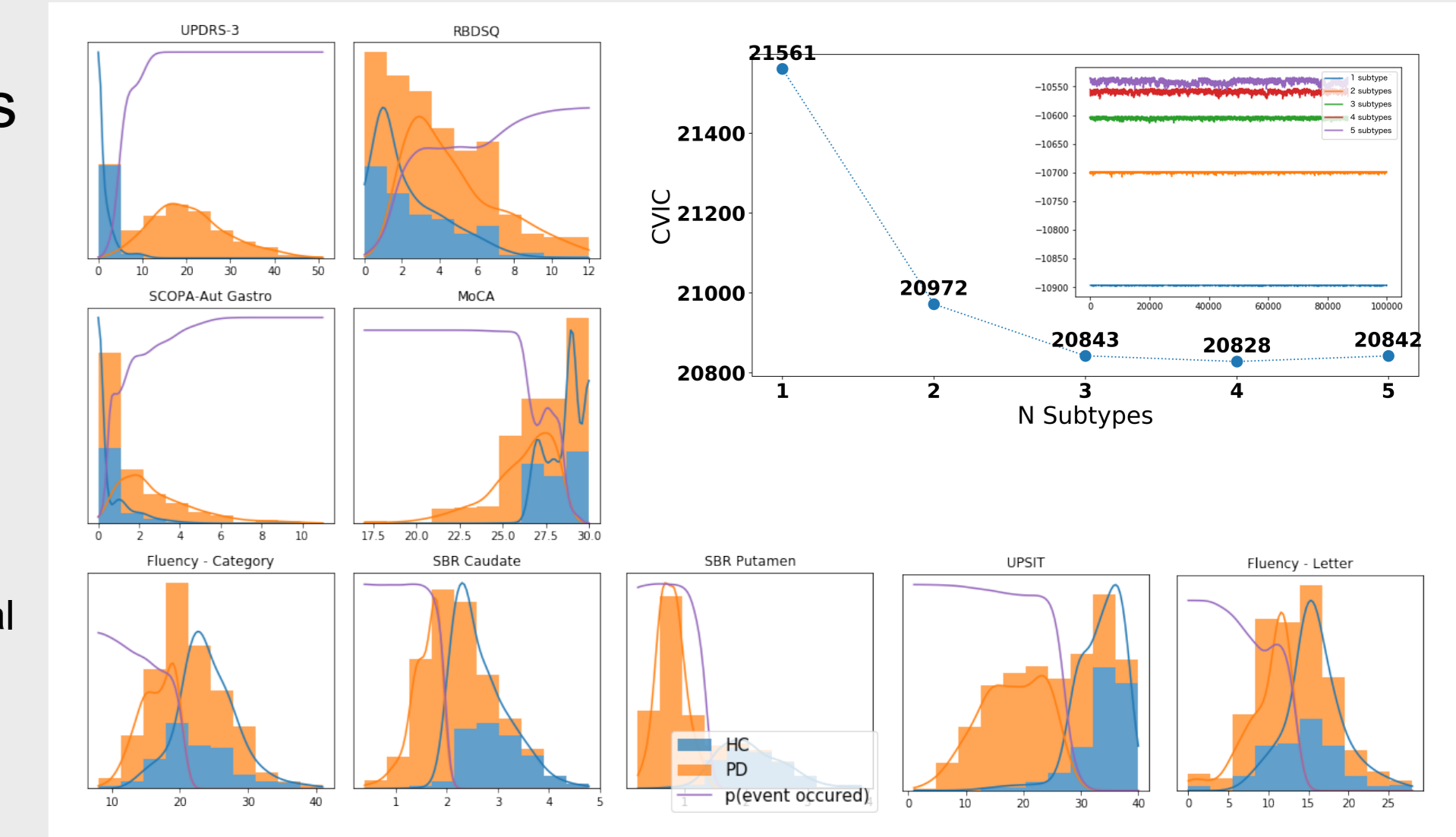
[ppmi-info.org](http://ppmi-info.org)

## Results: subtypes of Parkinson's

**Figure 1.** KDE Mixture Models for p(Event) and

Cross-Validation

UPDRS-3 = motor symptoms  
 RBDSQ = sleep problems  
 SCOPA Gastro = gastrointestinal  
 MoCA = cognition  
 SBR = dopamine deficiency  
 UPSIT = olfactory problems  
 HC = healthy controls

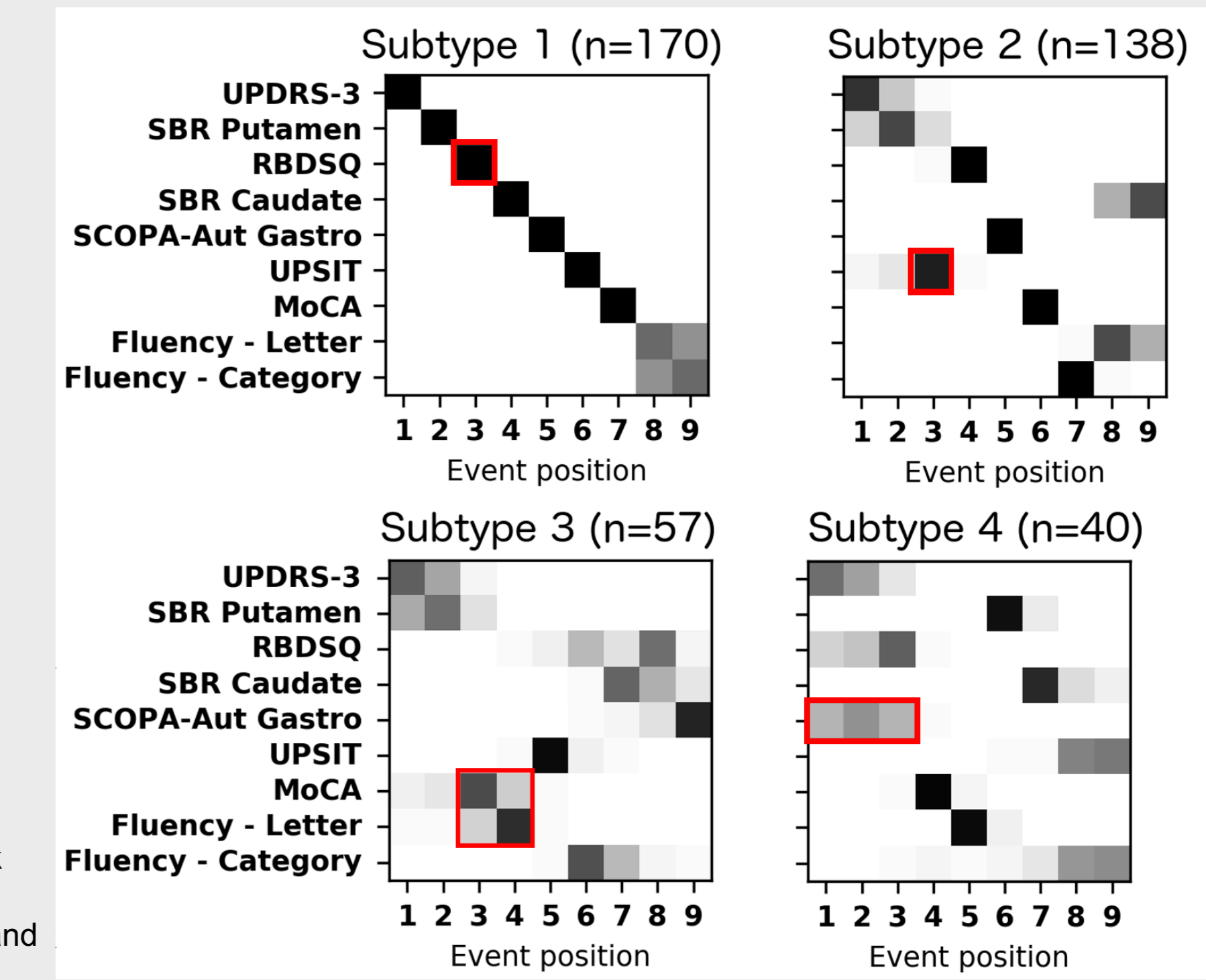


**Figure 2.** PD progression subtypes:

1. Sleep
2. Smell
3. Cognition
4. Gastrointestinal (SWEDD?)

Key observations:

- a) dopamine deficiency always putamen before caudate, and is usually early;
- b) when cognition is involved early, letter fluency is consistently before category fluency;
- c) sleep problems early in 3 subtypes.
- d) CVIC: is it 3/4/5 subtypes? => motivates methods-development work on improving model parsimony, e.g., explicitly hierarchical subtypes that split and merge as appropriate.



## Contributions

- Parkinson's disease understanding: fine-grained sequences of subtypes
- New, flexible KDE mixture modelling

## Considerations and future work

**Method:**

- Improve model parsimony

**Application:**

- Validate in prodromal cohorts
- Other markers:
  - MRI
  - Vision

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