

AI application in Medicine

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Lecturer Experimental Psychology

Background: AI in medicine

Applications of AI techniques in healthcare have skyrocketed over the past decade

- E.g. AI tools to develop new medicines
- AI tools to analyze wearables data
- AI tools to diagnose patients
- AI tools to investigate (neuro)biological underpinnings of disorders

Background: methods today

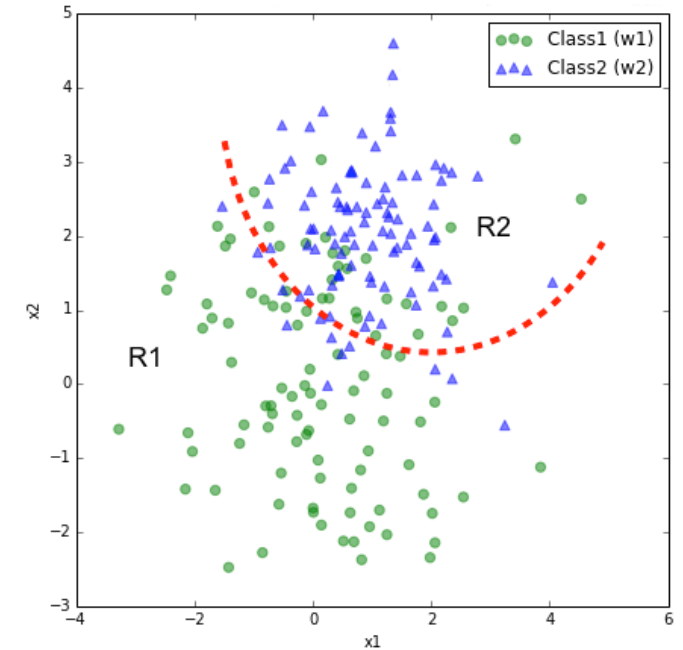
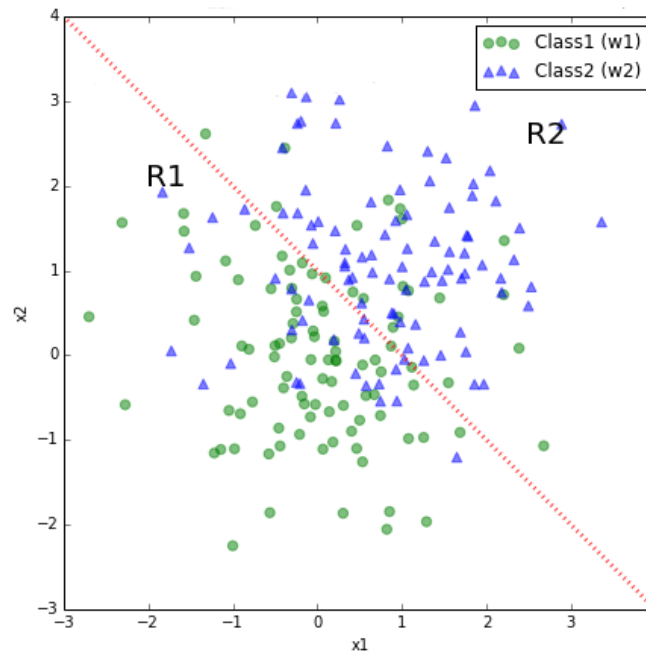
Classification & Stratification

- **Diagnosis:** Classification of subjects: patient or not? -> supervised
- **Etiology:** Stratification of patients: data driven subgroups -> unsupervised

Background: methods today

Classification & Stratification

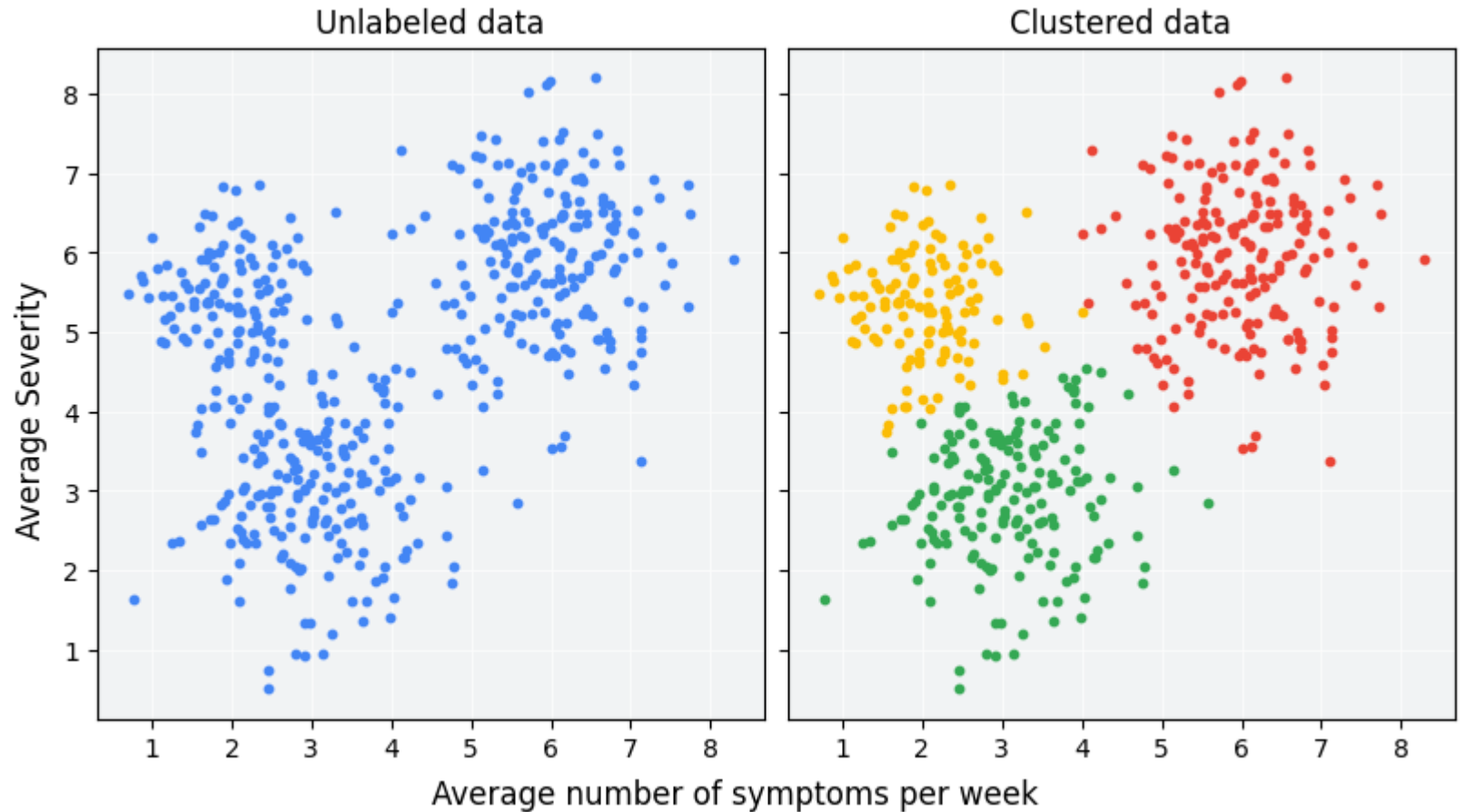
- Classification: using ML techniques to inform you which features are most predictive of a-priori group labels



Background: methods today

Classification & Stratification

- Stratification: using ML techniques to identify 'hidden' or underlying subgroups within a population



Background: topics for today

Three specific topics as examples












- Topic 1: Classification of ADHD patients based on fMRI patterns
- Topic 2: Predicting COVID-19 cases and deaths based on demographics
- Topic 3: Stratifying subjects with ASD based on their brain structure

Topic 1:

Classification of ADHD patients based on fMRI patterns

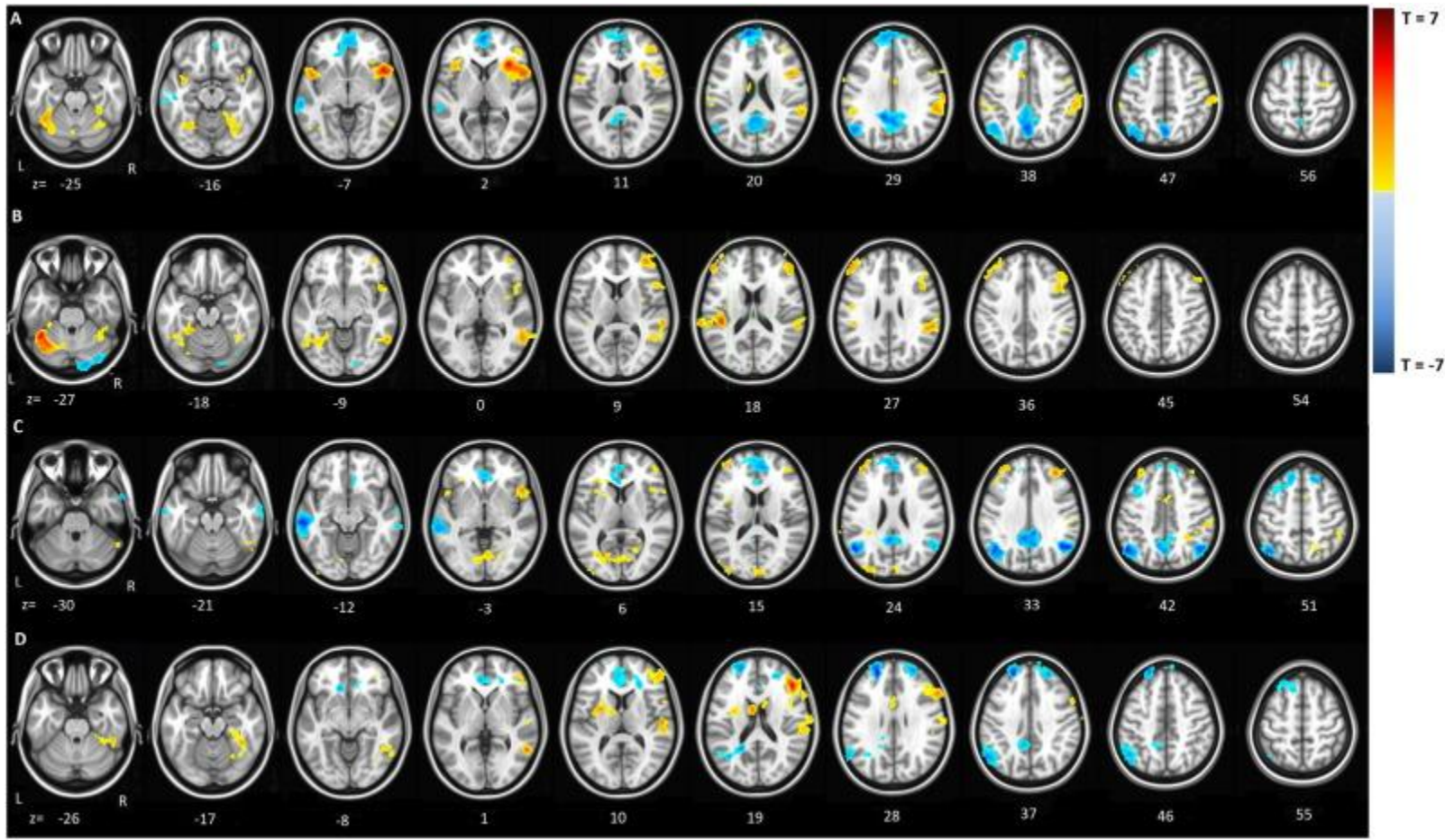
- Data: fMRI activation maps during an Inhibition task
- Classifier: Gaussian Process Classifier
- Outcome: Patient diagnosis

Quantifying patterns of brain activity:
Distinguishing unaffected siblings from
participants with ADHD and healthy
individuals

[Thomas Wolfers](#)^{a b}  , [Daan van Rooij](#)^b , [Jaap Oosterlaan](#)^c ,
[Dirk Heslenfeld](#)^c , [Catharina A. Hartman](#)^d , [Pieter J. Hoekstra](#)^d ,
[Christian F. Beckmann](#)^{b e f} , [Barbara Franke](#)^{a g} , [Jan K. Buitelaar](#)^{e h} ,
[Andre F. Marquand](#)^{b i} 

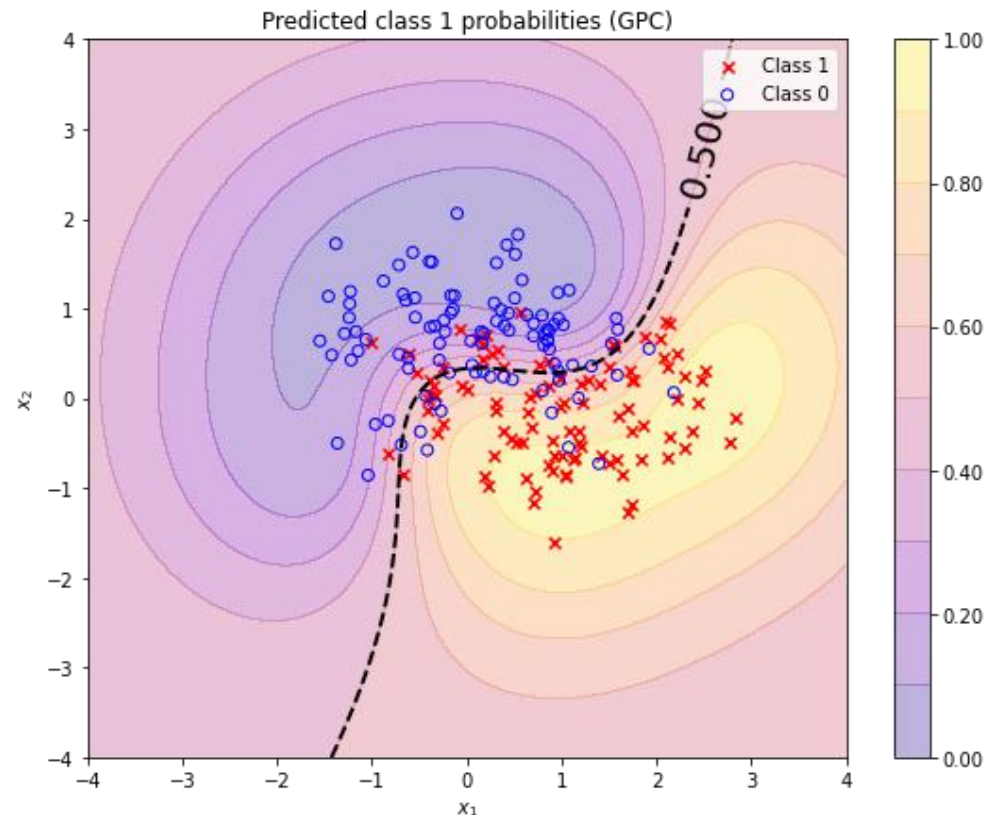
Topic 1: Classification of ADHD patients based on fMRI patterns

- Data: fMRI activation maps during an Inhibition task
- Classifier: Gaussian Process Classifier



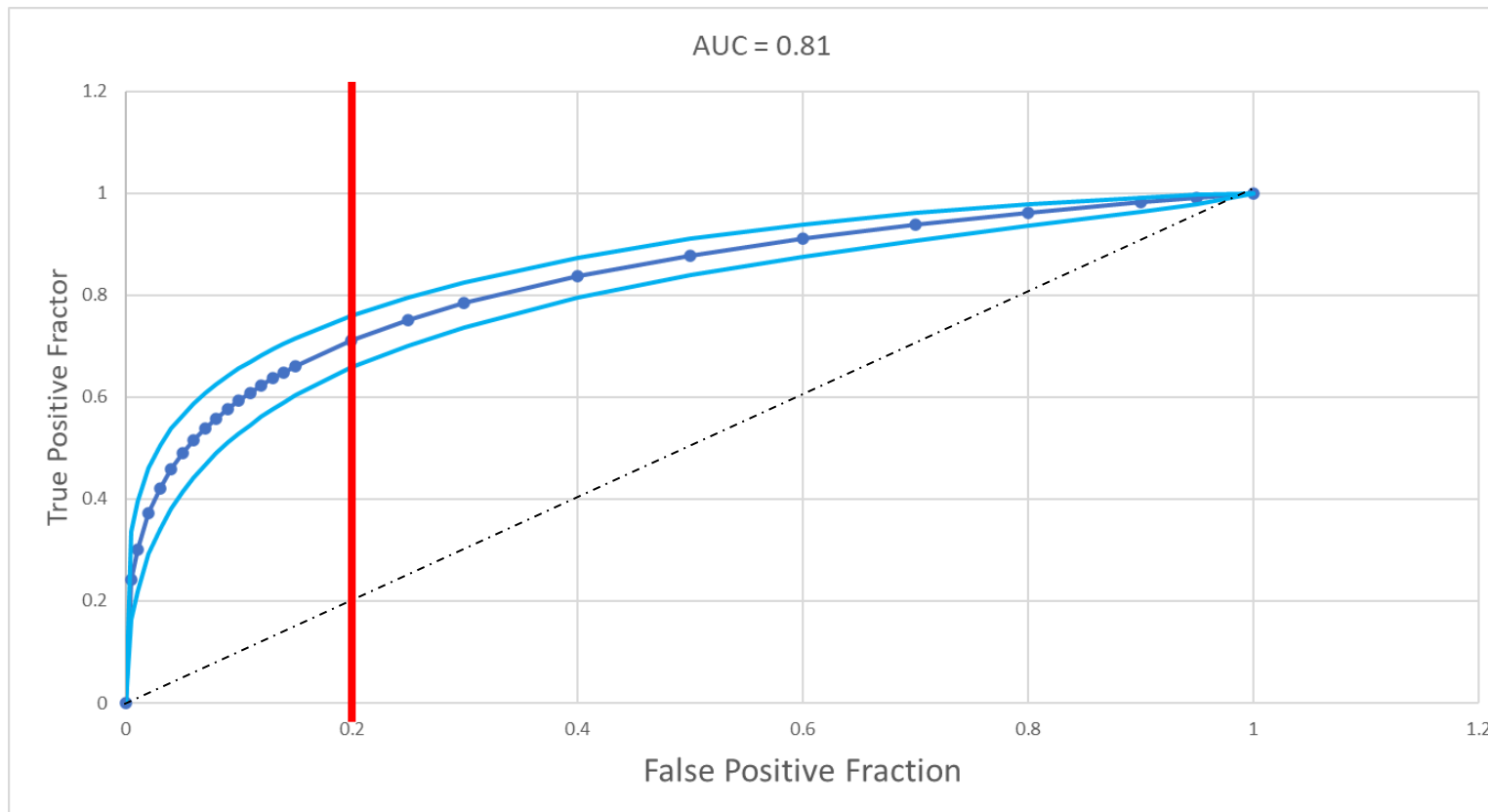
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Topic 1: Classification of ADHD patients based on fMRI patterns

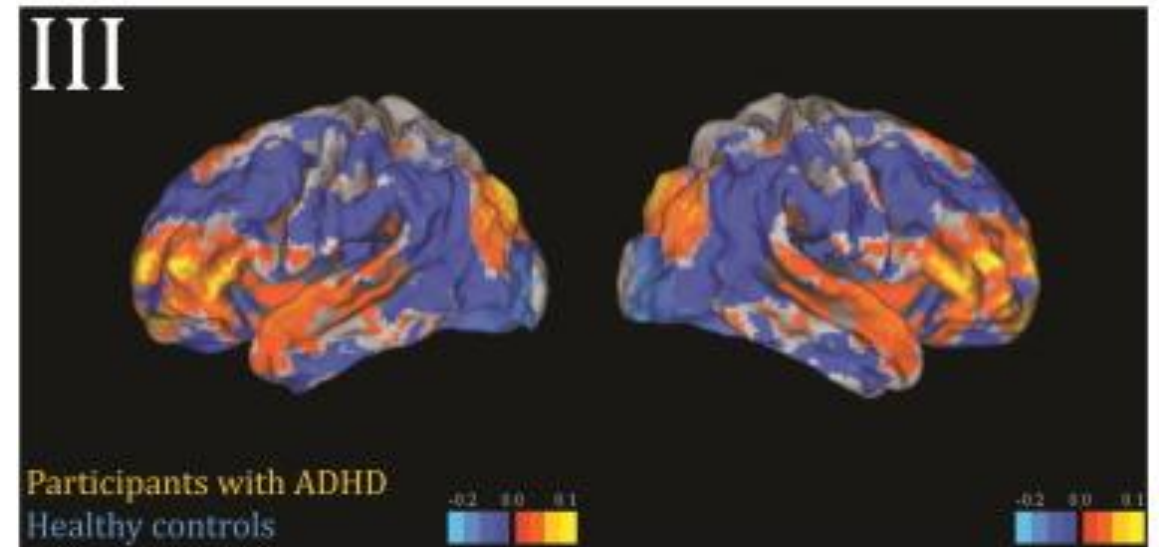
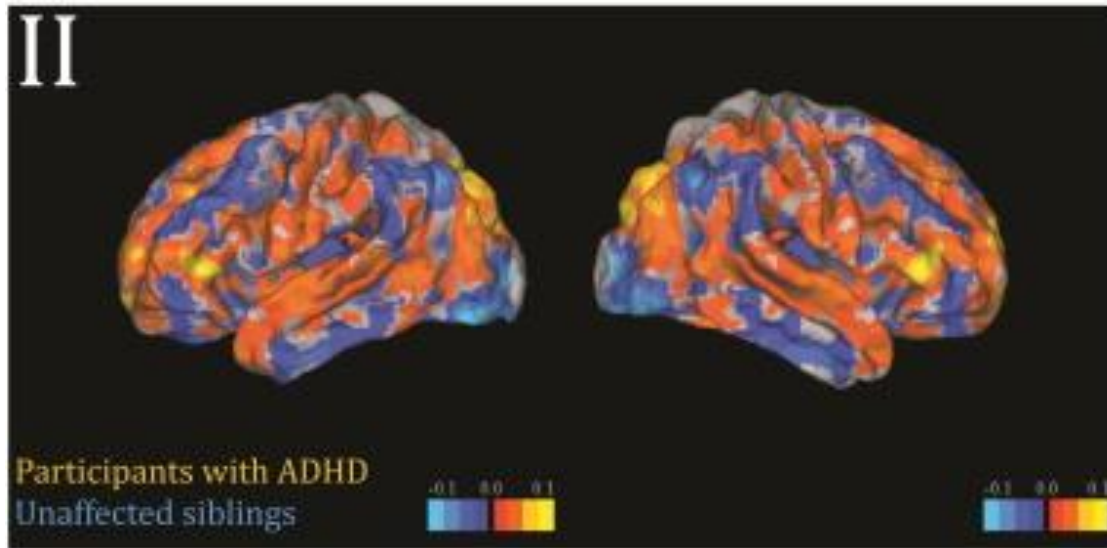
- Data: fMRI activation maps during an Inhibition task
- Classifier: Gaussian Process Classifier
- Outcome: Patient diagnosis + feature weight map



Number of Cases	700
Number Correct	540
Accuracy	77.10%
Sensitivity	75%
Specificity	80%
Pos Cases Missed	100
Neg Cases Missed	60
Fitted ROC Area	0.827
Empiric ROC Area	0.819

Topic 1: Classification of ADHD patients based on fMRI patterns

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Topic 1:

Classification of patients based on fMRI patterns

Promises:

- Better classification of patients
- More insight into neurobiology of disorders
- More insight into heterogeneity between subjects



Risks:

- Determinism/wrongful interpretation
- Inaccurate predictions
- Malignant use (e.g. by insurance companies)

Topic 2: Predicting COVID-19 cases and deaths based on demographics

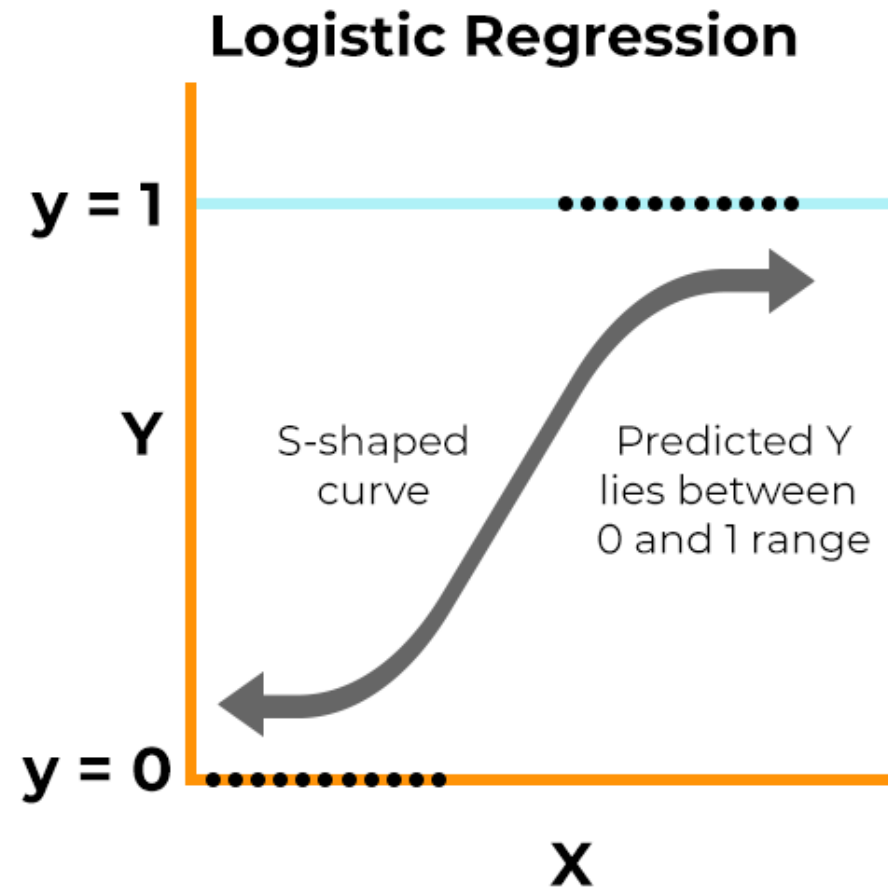
- Data: COVID-19 clinical data + population demographics
- Methods: logistic regression
- Outcome: odds-ratios/risk for COVID-19 cases & deaths (in a community)

Identifying novel factors associated with COVID-19 transmission and fatality using the machine learning approach

Mengyuan Li^{a b}, Zhilan Zhang^{a b}, Wenxiu Cao^{a b}, Yijing Liu^c, Beibei Du^c,
Canping Chen^{a b}, Qian Liu^{a b}, Md. Nazim Uddin^{a b}, Shanmei Jiang^{a b},
Cai Chen^d, Yue Zhang^{e f g}, Xiaosheng Wang^{a b}  

Topic 2: Predicting COVID-19 cases and deaths based on demographics

- Methods: logistic regression

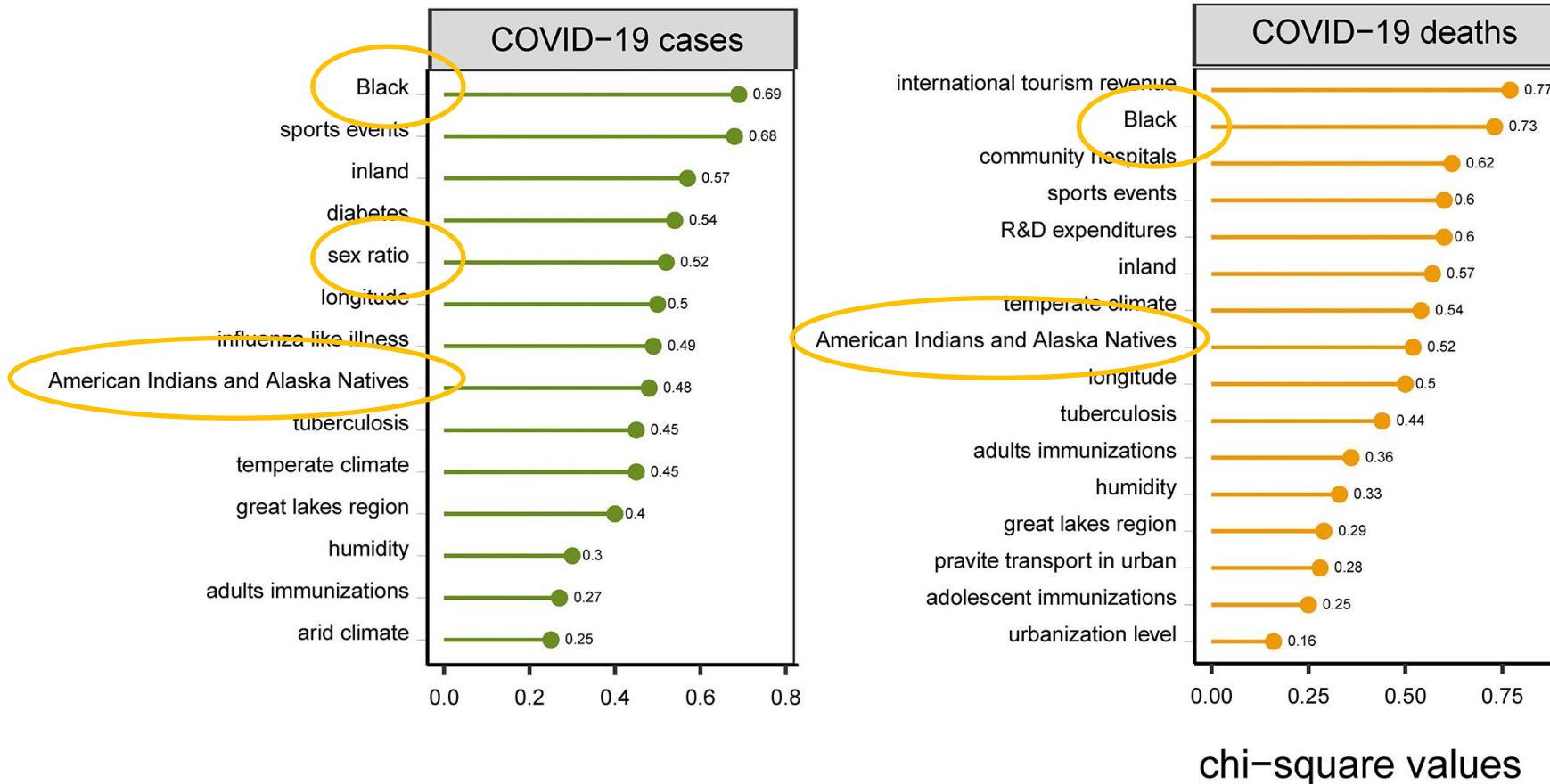


Topic 2: Predicting COVID-19 cases and deaths based on demographics

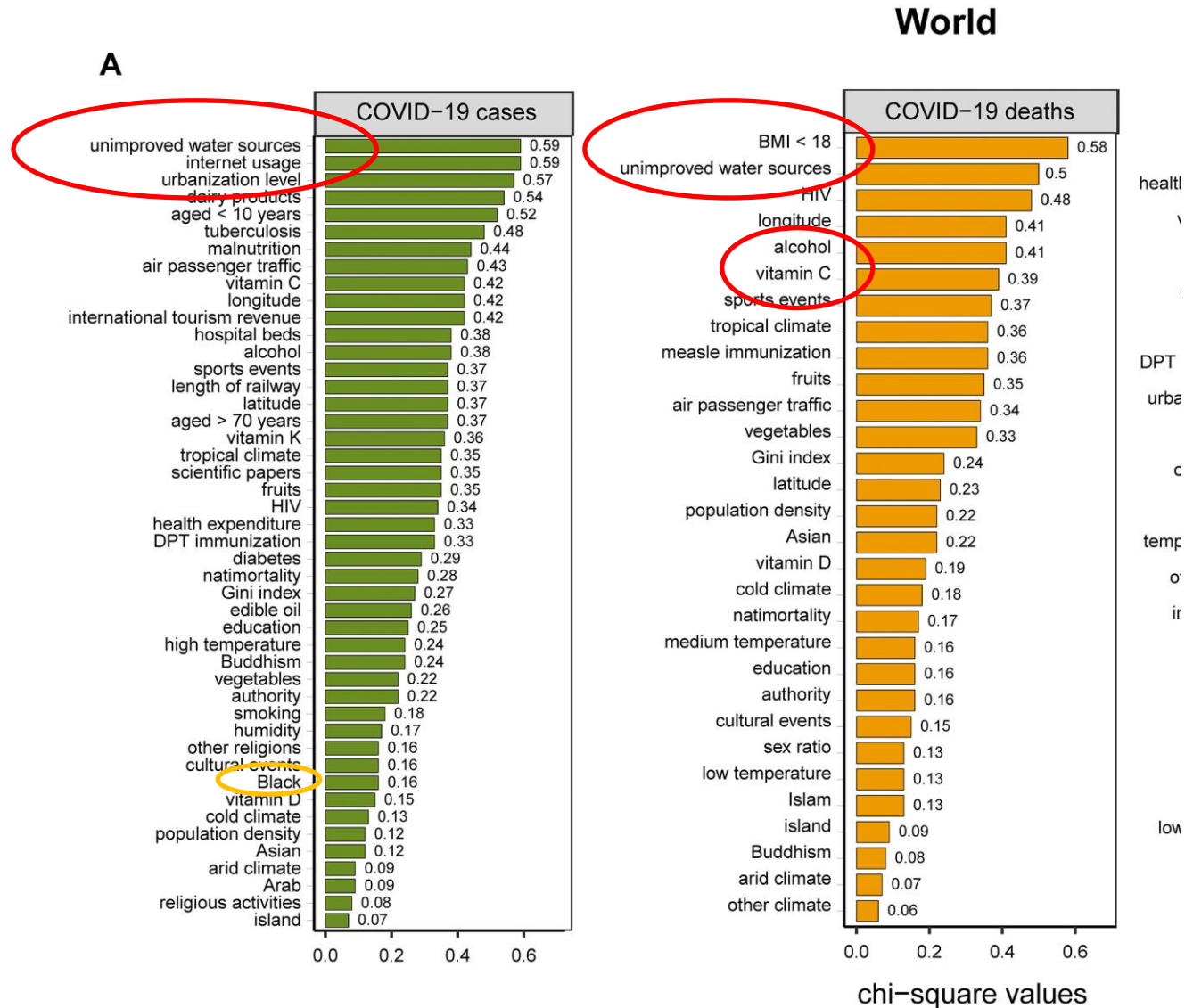
- Outcome: odds-ratios/risk for COVID-19

B

U.S.



Topic 2: Predicting COVID-19 cases and deaths based on demographics



Topic 2: Predicting COVID-19 cases and deaths based on demographics

Promises:

- Better understanding of COVID-19 spread
- Targeted interventions in at-risk populations
- Improvement of health care system

Risks:

- Discrimination/ inequality
- Wrongful attribution of causality (black-box)
- Wrongful biological determinism
- 'what you put in is what you get out'

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Topic 3:

Stratifying subjects with ASD based on their brain structure

- Data: structural morphometry of 53 brain segments
- Methods: normative modeling + spectral clustering
- Outcome: clusters of patients + clinical profiles

[nature](#) > [translational psychiatry](#) > [articles](#) > [article](#)

Article | [Open Access](#) | [Published: 06 November 2020](#)

Fractionating autism based on neuroanatomical normative modeling

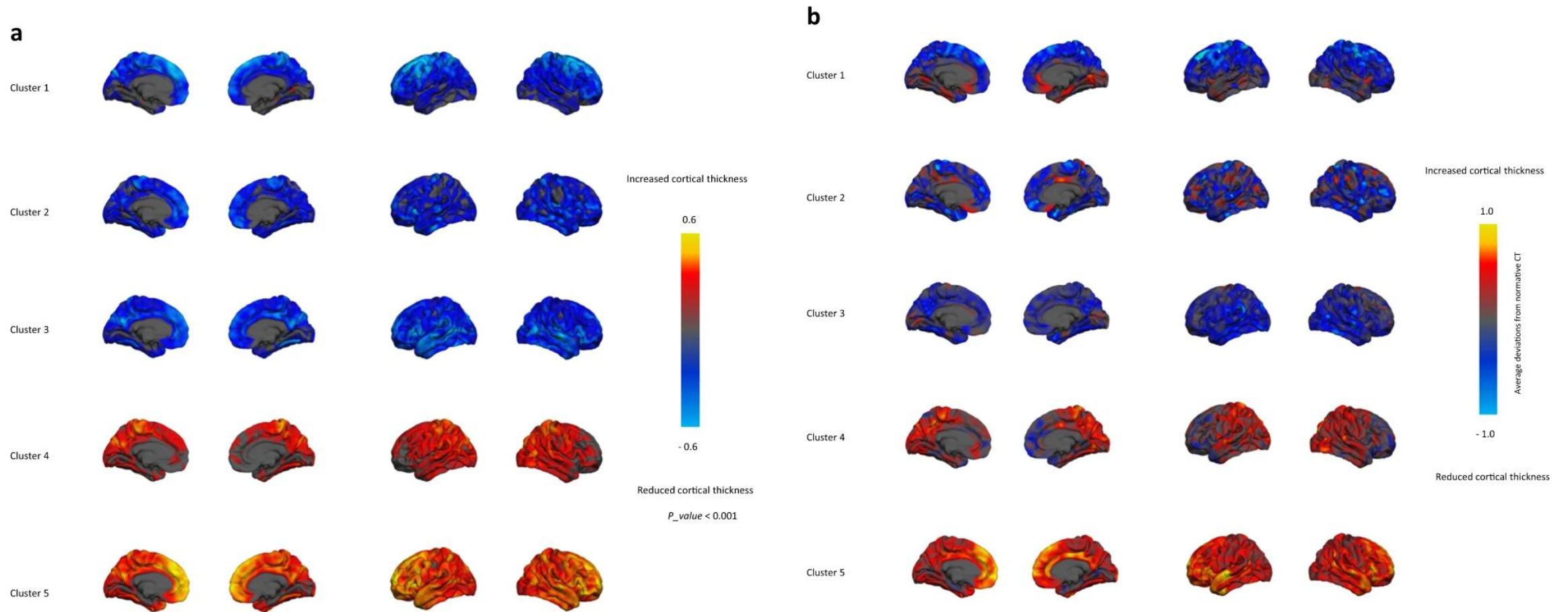
[Mariam Zabihi](#) , [Dorothea L. Floris](#), [Seyed Mostafa Kia](#), [Thomas Wolfers](#), [Julian Tillmann](#), [Alberto Llera Arenas](#), [Carolin Moessnang](#), [Tobias Banaschewski](#), [Rosemary Holt](#), [Simon Baron-Cohen](#), [Eva Loth](#), [Tony Charman](#), [Thomas Bourgeron](#), [Declan Murphy](#), [Christine Ecker](#), [Jan K. Buitelaar](#), [Christian F. Beckmann](#), [Andre Marquand](#) & [The EU-AIMS LEAP Group](#)

[Translational Psychiatry](#) **10**, Article number: 384 (2020) | [Cite this article](#)

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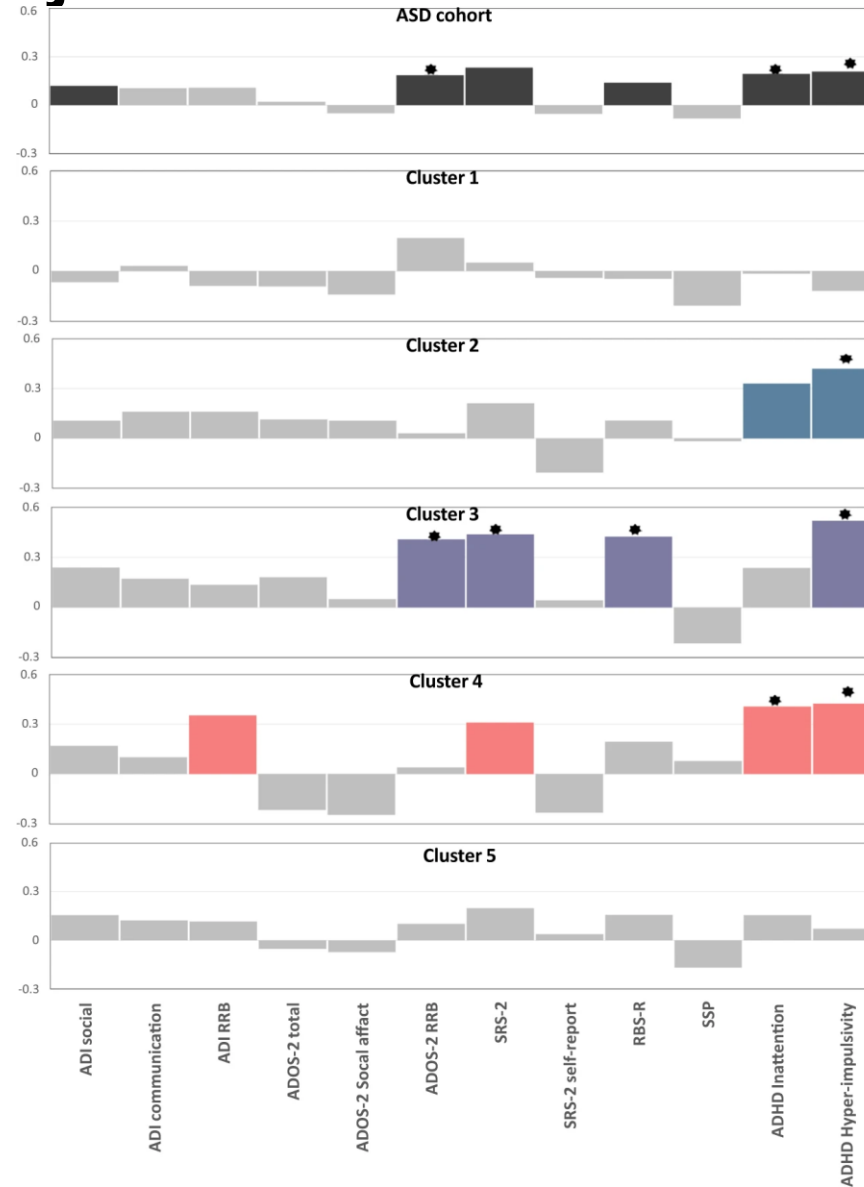
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- Outcome: clusters of patients + clinical profiles



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Repetitive Behavior Scale- Revised (RBS-R)

ADHD_ Hyperactive/ impulsivity

ASD cohort
N=249



ASD cohort
N=253



Cluster 3
N=51



Cluster 2
N=40



0.15  0.65
Correlation of atypically index with RBS

Short Sensory Profile (SSP)

Cluster 3
N=53



0.15  0.60
Correlation of atypically index with ADHD- hyperactivity/ impulsivity

Topic 3: Stratifying subjects with ASD based on their brain structure

Promises:

- More insight into heterogeneity between subjects
- Targeted interventions based on subject profile

Risks:

- Statistical spurious results

Conclusions:

- Many different applications for AI use in medical field
- More data, more data access, better AI methods
- Applications can have great benefits to science and health, but also risk
- Understanding of the AI methods necessary to mitigate risks



Utrecht University

Security in Open Societies

Future of Work

13-5-2026

Bottom-up Initiatives for Societal Change

Open Cities

Behaviour and Institutions

Transitions and well-being

Democracy and good governance

Fair Transitions

IOS-Themes: 15 platforms

Contesting Governance

Longtermism and Institutional Change

Futures of Democracy

The Transactional State as an Institution for Good

Markets and Corporations

Gender, Diversity and Global Justice

Openness challenged: the university at risk?

In/Equality

Equality and diversity