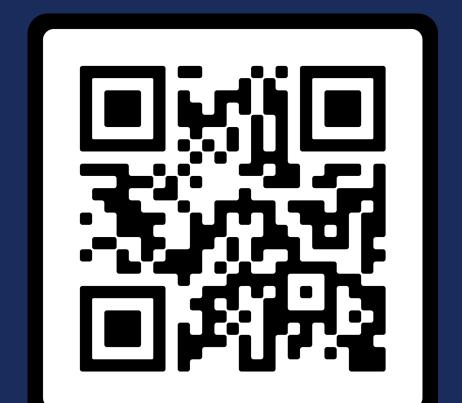
Development of a novel transdiagnostic predictor of psychiatric hospitalization using early trajectories of the Clinical Global Impression Scale

Maxime Taquet^{1,2}, Kira Griffiths³, Emily OC Palmer³, Sheryl Ker³, Christian Liman³, Soon Nan Wee³, Mayowa Oyesanya³, Scott H Kollins³,⁴, Rashmi Patel⁵* 1. Department of Psychiatry, University of Oxford, Oxford, UK., 2. Oxford Health NHS Foundation Trust, Oxford, UK., 3. Holmusk Europe Ltd., London, UK., 4. Department of Psychiatry and Behavioral Sciences, Duke University School of Medicine, Durham, North Carolina, USA., 5. Department of Psychological Medicine, King's College London, Institute of Psychiatry Psychology and Neuroscience, London, UK.



BACKGROUND

- Existing predictors of psychiatric hospitalization are often limited to specific diagnoses or clinical scenarios.
- The Clinical Global Impression Severity (CGI-S) is a 7-point measurements of illness severity¹ that can be used across psychiatric diagnoses and is frequently recorded in routine clinical practice.
- This study proposed an operationalization of clinical instability based on individual-level fluctuation in Clinical Global Impression – Severity (CGI-S) scores.
- We then investigated whether early instability of CGI-S scores predict risk of psychiatric hospitalization over a 6 month-follow-up period.

METHOD

• This retrospective cohort study using NeuroBlu² version 21R1 (Figure 1) a realworld data repository containing de-identified electronic health record (EHR) data from US mental healthcare providers.

Inclusion criteria

- An ICD-9 or ICD-10 code of major depressive disorder (MDD), bipolar disorder, generalized anxiety disorder (GAD), post-traumatic stress disorder (PTSD), schizophrenia or schizoaffective disorder (SCZ), attention deficit hyperactivity disorder (ADHD), or personality disorder (PD).
- At least 5 recorded CGI-S scores within a 2-month period, termed the index period.

Exclusion criteria

Record to hospitalization before or within the index period.

Statistical analysis

- The relationship between clinical instability and risk of psychiatric hospitalization was investigated using Cox regression.
- The primary outcome was any inpatient visit within 6-months following the index period (Figure 2).



OPERATIONALIZING CLINICAL INSTABILITY

- Clinical instability was the exposure of interest and this was defined as the generalized root mean squared subsequent differences (RMSSD).
- The root mean squared subsequent differences (RMSSD) is a measure of timeseries instability and has been previously used in psychiatry research.^{3,4}
- The conventional RMSSD assumes that all recorded values are equally separated in time. This is often not the case in EHR data, where the time between subsequent clinical encounters may vary.
- To account for differences in intervals between subsequent CGI-S measurements, the RMSSD was generalized.

The RMSSD was generalized as follows:

tRMSSD =
$$\sqrt{\frac{1}{N} \sum_{i=1}^{N-1} \left(\frac{C_{i+1} - C_i}{t_{i+1} - t_i} \right)^2}$$

where C_i (i=1,...,N) are the N measurements of CGI-S and t_i (i=1,...,N) are the timestamps at which measurements were recorded, so that t_{i+1} - t_i is the difference in time between two subsequent measurements. This generalization boils down to the original RMSSD if all t_{i+1} - t_i are equal.

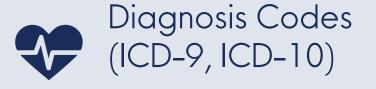
Neuro Blu THE EVIDENCE BLUEPRINT 80+ million 1.3M+ 20+ years Physical health data (Weight, Height, BP) rows of patient Patients Longitudinal Patient count by disease Substance-related disorders Major depressive disorder 347K Bipolar disorders 184K ADHD 115K Adjustment disorders 100K PTSD 86K Generalized anxiety disorder **80**K Personality disorders — 65K

Conduct disorders — 64K

Schizophrenia — 61K

Panic disorder = 27K

Structured Data







Patient Demographics Demographi

Semi-Structured Data

Clinical notes (intake notes, progress notes etc.) generated at every visit with rich information about the patient

Mental Status Examination (MSE)

Categorized notes on patient's function, appearance and



External Stressors Social, relational and occupational events that may affect the patient's mental health

RESULTS

- A total of 36,914 patients were included (mean [SD] age: 29.7 [17.5] years; 57.3% female).
- The median follow-up time was 180 days; interquartile range = 101-180 days.
- Clinical instability was associated with increased risk of psychiatric hospitalization (hazard ratio: 1.09, 95% Cl 1.07-1.10, p<0.001). There was no violation of the proportionality assumption (Chi²=0.62, P=0.43).
- These associations were consistent across all psychiatric diagnoses. The association was more marked among patients with ADHD and GAD, for whom an increase in clinical instability by 1 SD was associated with an 11% increased risk of hospitalization within the next 6 months (P < 0.0001). The association was weaker (although still statistically significant) for BD (HR 1.05, 95% CI 1.006-1.09, P=0.025) and SCZ (HR 1.07, 95% CI 1.004-1.14, P=0.038). Associations were also consistent across gender and age groups.

CONCLUSION

- We develop a novel measure of clinical instability using fluctuation of consecutive CGI-S measurements.
- Clinical instability may serve as a transdiagnostic clinical phenotype that provides information complementary to clinical severity and could improve outcome prediction in real-world clinical practice.

Limitations

• Limited definition of hospital admission: we could not distinguish between voluntary and involuntary. Further, only first admissions were captured and therefore we do not know the impact of clinical instability on recurrent hospital admission.

Conflicts of Interest: MT is a consultant for Holmusk inc. EOCP, KG, MO and CL are current employees of Holmusk Inc. SNW, SK and RP are previous employees of Holmusk Inc.

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