



Psychiatric Intensive Care &



Professor Faisil Sethi
MA (Cantab) MBBS MScDIC LLM DipStat FRCPsych
Chief Medical Officer, Dorset HealthCare University NHS FT
& Bournemouth University

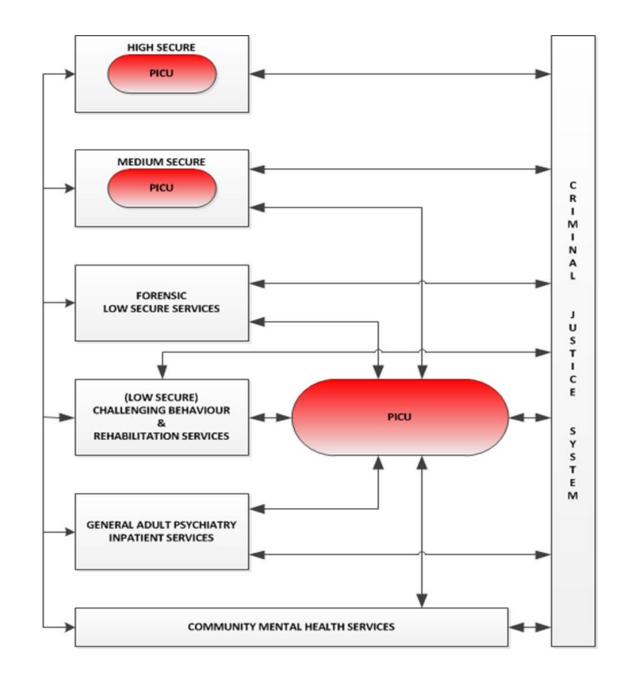
Leadership

CONTENT

- ☐ RESTRICTIVE INTERVENTIONS
- ☐ RAPID TRANQUILLISATION
- ☐ SECLUSION PRACTICE
- ☐ ART & MENTAL HEALTH
- ☐ DEMAND, CAPACITY & FLOW
- ☐ PANDEMIC & PICU
- ☐ VUCA

The Essence of PICUs

- Fast paced and high intensity.
- Immediacy of response.
- Acute disturbance of multiple aetiology.
- Multidisciplinary.
- Dynamic.
- Leadership at all levels.
- Treatment interventions reduce risk and improve clinical state.
- Innovative in approach.



Restrictive Interventions

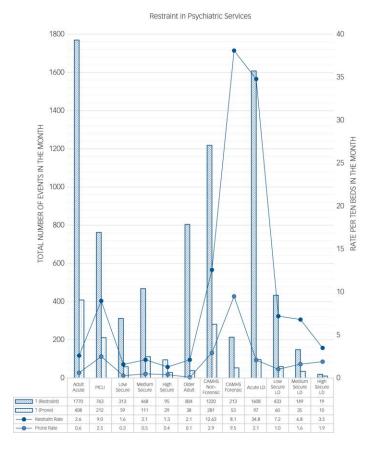


The British Journal of Psychiatry (2018) 212, 137–141. doi: 10.1192/bjp.2017.31

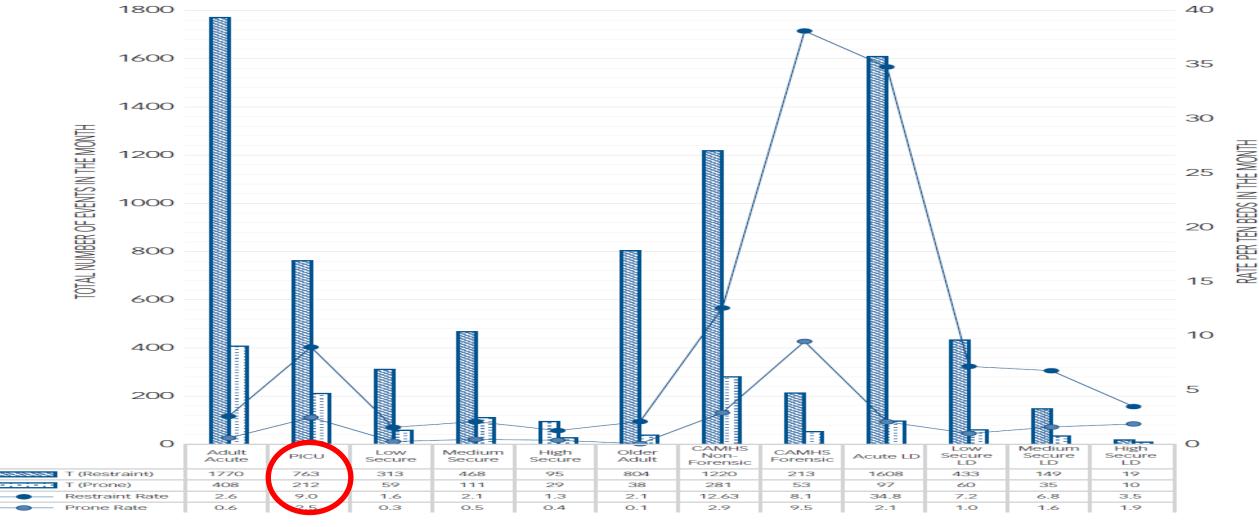
Analysis

Restraint in mental health settings: is it time to declare a position?

Faisil Sethi, John Parkes, Eric Baskind, Brodie Paterson and Aileen O'Brien



Restraint in Psychiatric Services



Analysis

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The British Journal of Psychiatry (2018)
Page 1 of 5. doi: 10.1192/bjp.2017.31

Restrictive Interventions (2019)

Journal of Psychiatric Intensive Care Journal of Psychiatric Intensive Care, 16(1): 23–28 doi:10.20299/jpi.2019.016 Received 10 June 2019 | Accepted 25 September 2019

BRIEF REPORT

Sensory room in a psychiatric intensive care unit

Rebecca Davies, Kenneth Murphy, Faisil Sethi



Sensory Rooms & Sensory Based OT Treatments (De-escalation)

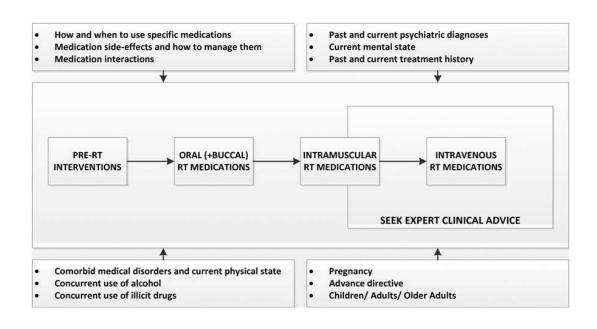
- Specially designed environment that offers a unique sensory experience
- Calming, de-escalating spaces but can also be immersive, interactive spaces
- Traditionally used in paediatric and learning disabilities
- Now used more often in adult psychiatric settings as an alternative method of de-escalation
- Support patients to improve skills in self-regulation of behaviour
- To potentially see a reduction in the use of restrictive interventions

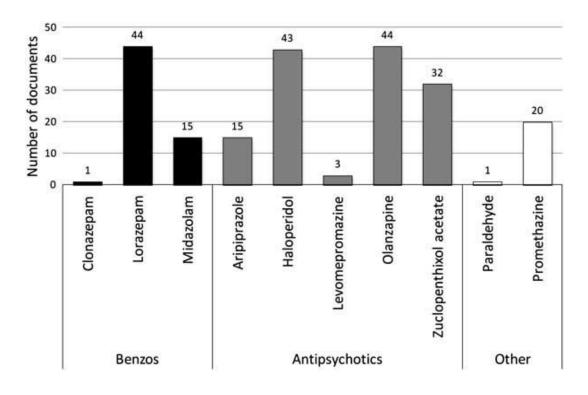






Rapid Tranquillisation (Innes & Sethi 2012)





Joint BAP NAPICU evidence-based consensus guidelines for the clinical management of acute disturbance: De-escalation and rapid tranquillisation

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Maxine X Patel^{1*}, Faisil N Sethi^{2*} With co-authors (in alphabetical order): Thomas RE Barnes³, Roland Dix⁴, Luiz Dratcu⁵, Bernard Fox⁶, Marina Garriga⁷, Julie C Haste⁸, Kai G Kahl⁹, Anne Lingford-Hughes¹⁰, Hamish McAllister-Williams^{11,12}, Aileen O'Brien¹³, Caroline Parker¹⁴, Brodie Paterson¹⁵, Carol Paton¹⁶, Sotiris Posporelis¹⁷, David M Taylor¹⁸, Eduard Vieta⁷, Birgit Völlm¹⁹, Charlotte Wilson-Jones²⁰ and Laura Woods²¹

Journal of Psychopharmacology **32**(6): 601-640.

PMID: 29882463

AIM: To review evidence and provide recommendations on de-escalation and medication



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1–39
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Joint BAP NAPICU evidence-based consensus guidelines for the clinical management of acute disturbance: deescalation and rapid tranquillisation

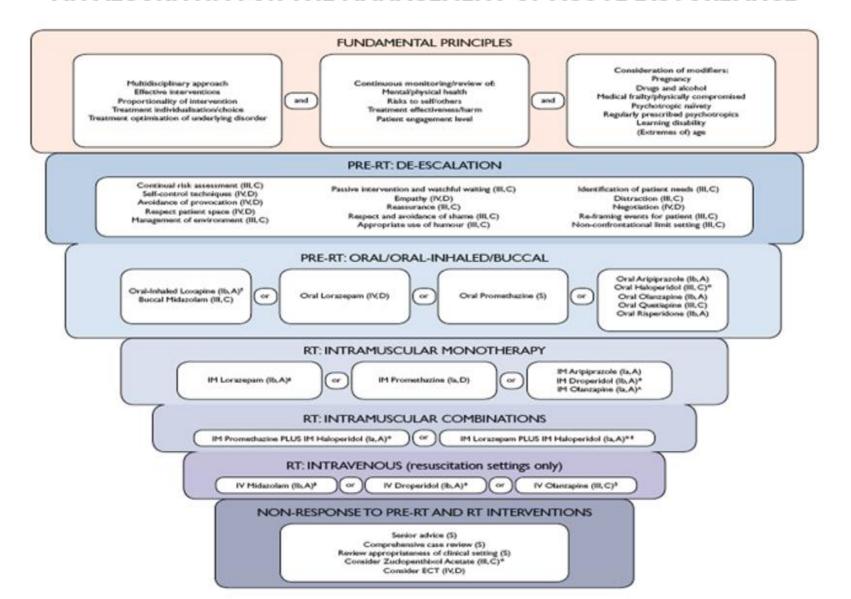
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Maxine X Patel^{1,*}, Faisil N Sethi^{2,*}, Thomas RE Barnes³, Roland Dix⁴, Luiz Dratcu², Bernard Fox⁵, Marina Garriga⁶, Julie C Haste⁷, Kai G Kahl⁸, Anne Lingford-Hughes^{3,9}, Hamish McAllister-Williams^{10,11}, Aileen O'Brien^{12,13}, Caroline Parker¹⁴, Brodie Paterson¹⁵, Carol Paton¹⁶, Sotiris Posporelis^{17,18}, David M Taylor¹⁷, Eduard Vieta⁶, Birgit Völlm¹⁹, Charlotte Wilson-Jones² and Laura Woods²⁰





AN ALGORITHM FOR THE MANAGEMENT OF ACUTE DISTURBANCE





SEVEN FUNDAMENTAL OVERARCHING PRINCIPLES

Multidisciplinary approach:

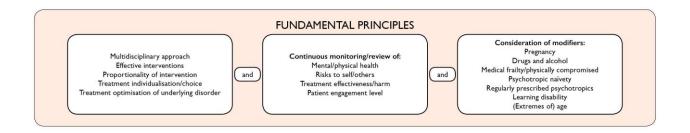
- aetiology of acute disturbance is complex and heterogeneous
- Mx: psychopharmacological, psychological, environmental and social interventions

Effective interventions:

 evidence base confirming that they increase positive outcomes and/or reduce negative outcomes (harm) of acute disturbance, in the immediate to short-term (minutes to hours)

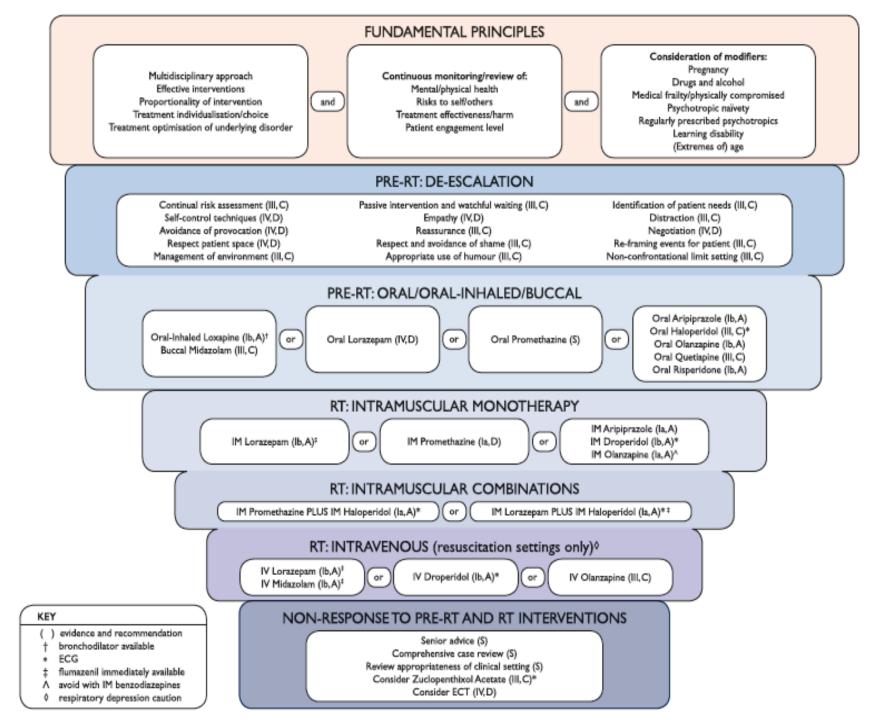
Proportionality of intervention:

- an intervention's associated restriction on the patient should be proportionate (i.e. not excessive) to the acute severity of the clinical risk posed by the acute disturbance
- least restrictive options available should always be considered first



- Treatment individualisation/choice:
 - consideration of patient specific factors (clinical, risk and choice related)
- Treatment optimisation of underlying disorder:
 - Interventions should be set in a context of the overarching goal of optimising the treatment of the underlying disorder
- Continuous monitoring/ review of:
 - (i) mental/physical health (iii) treatment effectiveness/harm (ii) risk to self/others (iv) patient engagement level
 - Risk is dynamic, and intervention selection needs to reflect this so that
 - the right intervention is used for the right scenario at the right time

- 2018: Step change in the management of Acute Disturbance
- Still gaps in evidence, but there is evidence to inform practice
- Oral / IM / IV recommendations
- Exact choice should be tailored to the patient



Seclusion Practice

Journal of Psychiatric Intensive Care Journal of Psychiatric Intensive Care, 13 (2): 83–91 doi:10.20299/jpi.2017.007 Received 18 August 2016 L Accepted 23 Morch 2012

Received 18 August 2016 | Accepted 23 March 2017 © NAPICU 2017

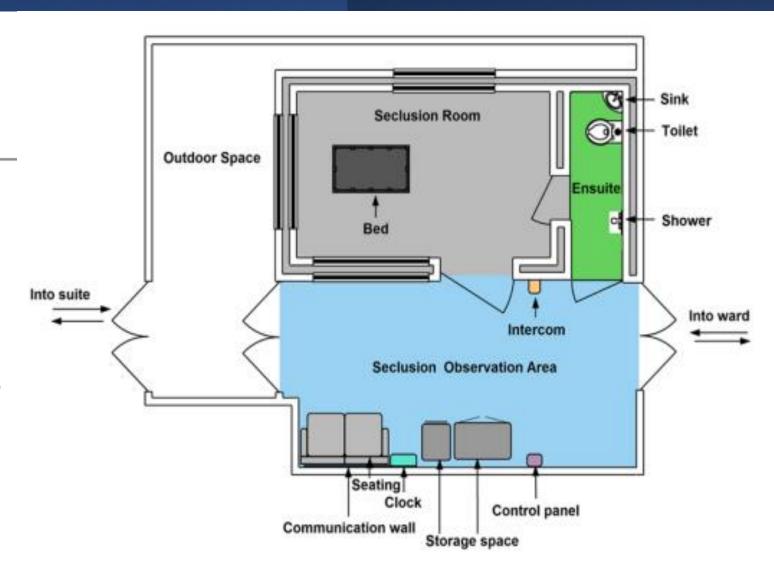
GRAPICO 2017

REVIEW ARTICLE

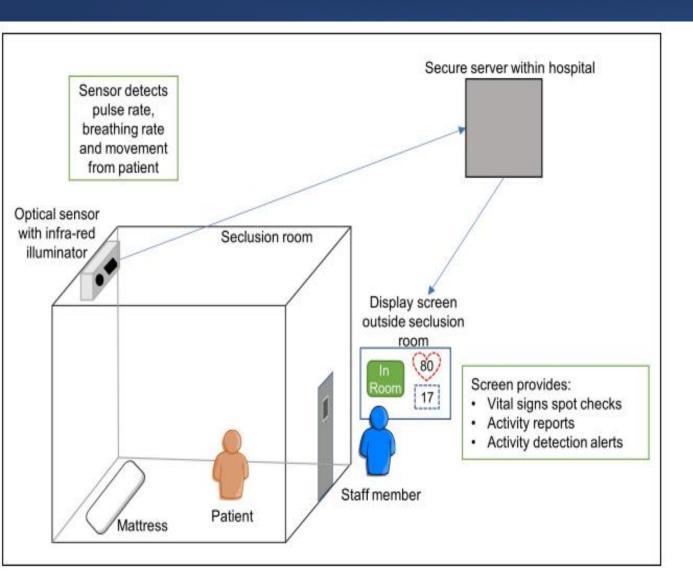
The function and design of seclusion rooms in clinical settings

Stephen J Kaar, Helen Walker, Faisil Sethi, Ronan McIvor

South London & Maudsley NHS Foundation Trust



Seclusion Practice (2020)



Journal of Psychiatric Intensive Care Journal of Psychiatric Intensive Care, 18 (1): 31–37 doi: 10.20299/jpi.2021.009 Received 25 March 2021 | Accepted 6 July 2021 © NAPICU 2022

BRIEF REPORT

Non-contact physical health monitoring in mental health seclusion

Hannah Clark¹, Allison Edwards¹, Rebecca Davies¹, Adenike Bolade¹, Rachael Leaton¹, Robert Rathouse¹, Marisa Easterling¹, Ronnie Adeduro¹, Matthew Green¹, Wellington Kapfunde¹, Oladapo Olawoyin¹, Kalliopi Vallianatou¹, Daniel Bayley², Oliver Gibson², Charlotte Wood², Faisil Sethi³

¹South London & Maudsley NHS Foundation Trust, UK; ²Oxehealth, Oxford, UK; ³Dorset HealthCare University NHS Foundation Trust

Challenges to physical health

Challenges in seclusion:

- Higher risk of physical health conditions
- Communication difficulties
- Rapid tranquilisation and restraint

National guidance

- RT 15min obs
- Seclusion 2 hourly obs

Difficulties

- Patient engagement
- Challenging seclusion entries
- Maximising therapeutic rest

Novel & innovative technology



- Optical sensor = camera + infrared illumination in secure housing in room
- Screen outside seclusion room
- Pulse & breathing rate measurements without disturbance
- World first medical device
- Data on patient movement

Breaking Barriers (2018)

Journal of Psychiatric Intensive Care

Journal of Psychiatric Intensive Care, 16(1): 15-21 doi:10.20299/jpi.2019.015 Received 10 January 2019 | Accepted 1 August 2019 © NAPICU 2020

BRIEF REPORT

Art and mental health in the women's psychiatric intensive care unit

Sophie Butler¹, Ronnie Adeduro¹, Rebecca Davies¹, Onyekachi Nwankwo¹, Niamh White², Timothy A Shaw², Luke Skelton¹, Geordan Shannon³, Emma Smale², Marie Corrigan³, Daria Martin⁴, Faisil Sethi¹

¹South London & Maudsley NHS Foundation Trust, Maudsley Hospital, London, UK; ²Hospital Rooms, UK; ³Global Health Disrupted, UK; ⁴The Ruskin School of Art, University of Oxford, UK



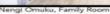






Aimee Mullins, Seating Area







Harold Offeh, TV Room







Tim A Shaw, Conference Room



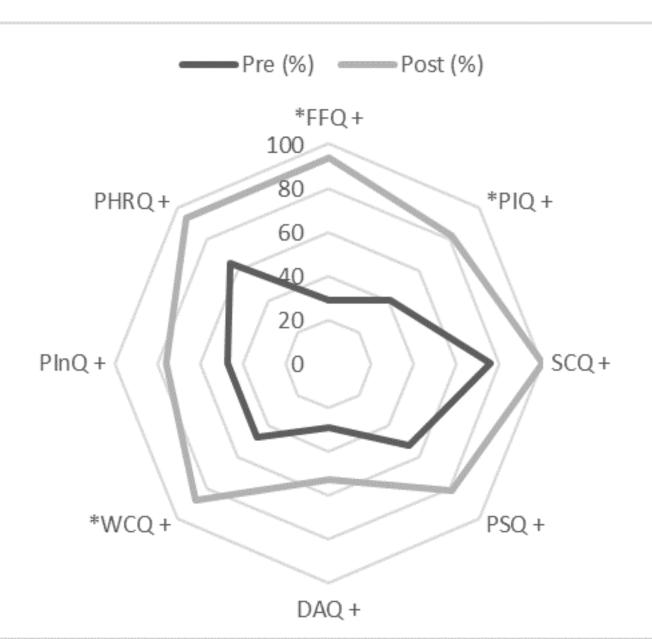








IMPACT & EVALUATION



- Engagement and Participation
- Service User
- Staff

| | Code | Question |
|---|------|---|
| (| FFQ | How likely are you to recommend our ward to friends and family if they needed similar care or treatment? |
| | PIQ | Do you feel involved in your care? |
| | SCQ | Are staff kind and caring? |
| | PSQ | Do you feel safe here? |
| | DAQ | Are there activities 7 days a week? |
| | WCQ | Do you think the ward is comfortable? |
| | PInQ | Do we treat you as an individual by considering your culture, spirituality, disability, gender, sexuality, age and ethnicity? |
| | PHRQ | Do you have hope that the care you are having from this ward will help you? |

Demand, Capacity & Flow







South London and Maudsley Will

Psychiatric Intensive Care Units:

patient types, admission lengths and bed numbers

S. Dye ¹, F. Sethi ², V. Koravangattu ³, E. Rose ³, L. Penfold ¹, M. Campbell ⁴, T. Kearney ⁵

¹ Norfolk and Suffolk NHS Foundation Trust ² Dorset Health Care University NHS Foundation Trust ³ South London and Maudsley NHS Foundation Trust ⁴ Central and North West London NHS Foundation Trust ⁵ NHS England and Improvement

INTRODUCTION

Patients spend varying lengths of time within a PICU, ranging from 1 day to over a year. Individual PICUs vary in capacity, but standards dictate that a ward's maximum size is limited to 14 beds.1

Health services aim to provide correct care in a timely fashion based upon patient needs. Classifying individual needs is distinct from diagnosing disorders or identifying underlying pathologies

Within mental health services in England and Wales, a specific tool has been developed to identify patient needs by separating patient presentations according to a specific assessment: the Mental Health Clustering Tool (MHCT),2 This mandatory assessment is based on the Health of the Nation Outcome Scale (HoNOS).3

It has been demonstrated that the MHCT does not distinguish patient need within a PICU and that a 'standard PICU admission' does not exhibit a typical clustering pattern.4 Instead, different identifiable clinical types of patients can be grouped by clinical need. Groups were associated with differing lengths of stay and discharge destinations. Discharge destination is something that the clinical team should be aware of early in a patient's admission (if not beforehand) and is a measure of

OBJECTIVES

Primary objective: Use statistical techniques to determine if patient typing by discharge destination can be a useful classification tool to help predict length of stay within PICUs.

Subsidiary objective: Determine if a modelling process using patient typing can help measure needed ward capacity.

MATERIALS & METHODS

Following a focus group discussion with clinicians, specific PICU patient types were developed to be used within this project. Patients are admitted from, and discharged to, the following mutually exclusive groups:

- 1. TYPICAL: Adult acute psychiatric inpatient services or community settings (including A&E or subsequent to Section 136 (being brought to a place of safety by
- 2. LONGER SECURE CARE: Care within a longer term secure environment (e.g. Medium or Low Secure Unit, 'Locked Rehabilitation' Unit)
- 3. OTHER: Other psychiatric settings (e.g. older adult wards, mother and baby units, another (normally out of area) PICU or other inpatient specialist services)



Four, 10-bedded PICUs were used: The lead NHS organisation (NSFT) approved the project as service evaluation and the University of Sheffield approved the project from

Data analysis was performed using R [code available upon request].5

Length of stay (LoS) is usually short (days). However, some 'complicated' patients may stay for more extended periods, and thus the LoS distribution is skewed (with the mean being greater than the median). In an attempt to solve this, admission length values were transformed logarithmically using natural logarithms.

Although the patient type categorisation was not precisely the same, data from Dye (2017) provided approximations for power analysis. A power analysis using these figures for a one-way ANOVA suggested that each group should have 25 patients for a power of 0.95 at a 5% significance level. For pragmatic reasons, attempts were made to collect data on up to the last 40 patients discharged from each patient-type group from each of the PICUs. This was limited to a retrospective period of 4 years.

40 Patient Type 1 patients (TYPICAL) from each unit were easily collected. From the other types, the number discharged over the maximum four-year time period

Apart from the individual PICU and patient type, other variables can influence LoS and its variation, but the number of variables was minimised as far as possible to make this a clinically useful project. These were:

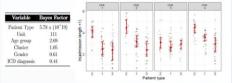
Patient Age Patient Gender Primary Diagnosis HoNoS Cluster (six categorical groupings) (ICD-10 chapter)

RESULTS

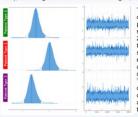
The number of records analysed within the four year period were:

| UNIT | Patient Type 1 | Patient Type 2 | Patient Type 3 | TOTAL |
|-------|----------------|----------------|----------------|-------|
| 1 | 40 | 22 | 38 | 100 |
| 2 | 40 | 20 | 28 | 88 |
| 3 | 40 | 23 | 17 | 80 |
| 4 | 40 | 23 | 37 | 100 |
| TOTAL | 160 | 88 | 100 | 368 |

Bayes factor is a ratio of the likelihood of observing the data given one model over the likelihood of observing the data given another. If it is greater than 1, then the first model is favoured, and the higher the number, the more that model is favoured (3-10:moderate evidence, >10: strong evidence). If the 'null' model is one with all parameters for variables collected being zero, each variable can be introduced while keeping the others at zero. Within this study, Bayes factors introduced in this manner show that patient type is the preferred variable. There is no evidence to support 'cluster' being used (Bayes factor of 1.04) and anecdotal evidence that the model with no parameters is better than that with the ICD(diagnosis) variable (Bayes factor of 0.44)!



The above figure reveals evidence of an interaction between unit and patient type variables: patient type 2 (LONGER LOW SECURE) individuals having shorter lengths of stay in Unit 3 compared to the other units. A parsimonious model for LoS would be one determined by patient type, unit and interaction between patient type and unit. When this is simulated over the units using 4 chains of 5000 iterations (each with a 'warm-up' of 1000), converting back to actual admission lengths, the following distributions are seen:



The left side of the figure gives densities of draws taken from the posterior distribution. This shows some separation of patient type 2. The right part shows trace plots that indicate good mixing of simulations and reasonable (but not excellent) model convergence: the caternillar plot is straight but the centre spans a large range).

To fully model ward capacity, the demand for PICU needs to be considered. Unfortunately, although admission rates were calculated, the number not admitted although

Needing admission could not be obtained. Despite this, the modelling process was performed (to demonstrate how it can be done). The number of patients turned away due to units being at capacity was modelled for each unit over a four-year period (using the R simmer package). Although these are bound to be under-estimates, the table below displays the simulated number of turn aways for each unit over 4 years:

| Unit | 0.25 quantile | Median | 0.75 quantile |
|------|---------------|--------|---------------|
| 1 | 2 | 4 | 7 |
| 2 | 4 | 8 | 13 |
| 3 | 10 | 1.4 | 19 |
| 4 | 2 | 4 | 8 |

CONCLUSIONS

Patient-typing contributed most to differences in admission length upon the PICUs studied (compared to age, sex, diagnosis and clustering). The unit to which a patient was admitted also influenced LoS. A model using both these variables and an interaction between them showed reasonable utility.

A method for modelling bed capacity based upon turn-away rates was introduced

REFERENCES

1. NAPICU, 2014. National Minimum Standards for Psychiatric Intensive Care in General Adult Services. National Association of Psychiatric Intensive Care Units.

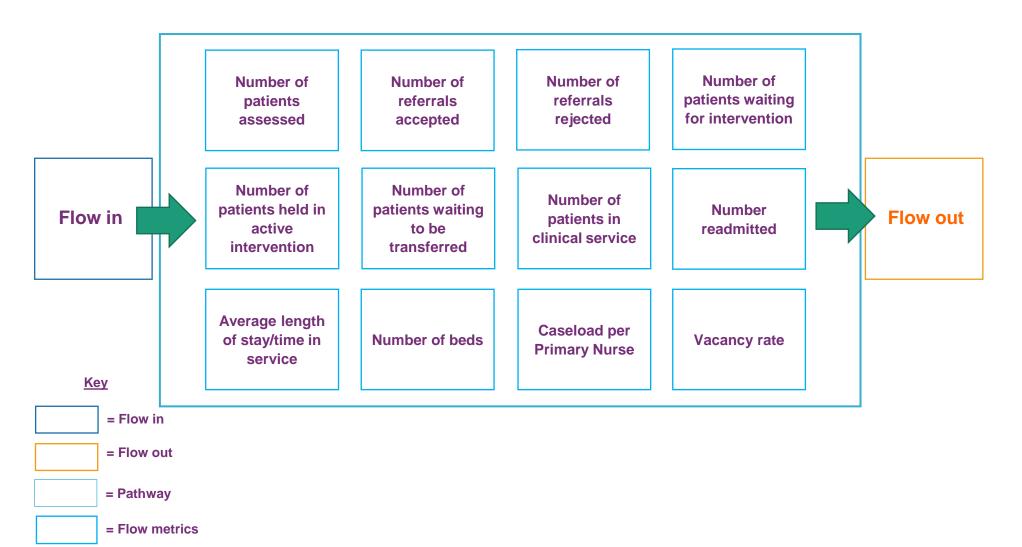
2. NHS England. 2016. Mental Health Clustering Booklet (V5.0 2016/17). NHS England

3. Wing, J., R. Curtis, and A. Beevor. 1999. "Health of the Nation Outcome Scales (HoNOS)." The British Journal of 4. Dye, S. 2017. "Can Mental Health Clusters Be Replaced by Patient Typing?" British Journal of Healthcare Management

5. R Core Team. 2021. R: A Language and Environment for Statistical Computing. Vienna, Austria: R Foundation for

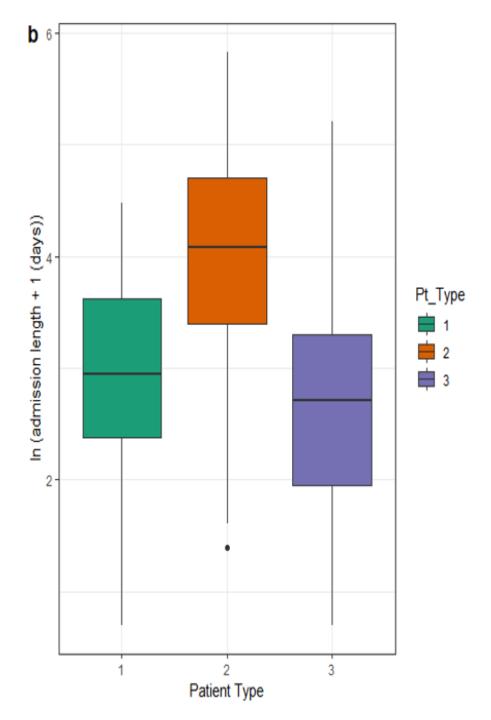
Which metrics support flow?

Example: Psychiatric Intensive Care Unit (PICU)



Models to Predict LOS in PICUs (Dye, Kearney, Sethi et al)

| Patient Type | Median (days) | |
|-----------------|------------------|--|
| Typical | 18.00 | |
| Secure Care | 58.00 | |
| Other | 14.00 | |



Pandemic and the PICU

Journal of Psychiatric Intensive Care Journal of Psychiatric Intensive Care, 17(1): 61–68 doi:10.20299/jpi.2020.022 Received 26 November 2020 | Accepted 11 December 2020 © NAPICU 2021 Journal of Psychiatric Intensive Care Journal of Psychiatric Intensive Care, 16 (2): 85–88 doi:10.20299/jpi.2020.008 Received 1 May 2020 | Accepted 11 May 2020 © NAPICU 2020 Journal of Psychiatric Intensive Care Journal of Psychiatric Intensive Care, 16 (2): 89–93 doi:10.20299/jpi.2020.009 Received 28 May 2020 | Accepted 9 June 2020 © NAPICU 2020

BRIEF REPORT

The psychiatric intensive care unit clinical model and COVID-19

Faisil Sethi¹, Luke Skelton², Lucy Blake², Elizabeth Rose², Stephen Dye³

¹Dorset HealthCare University NHS Foundation Trust; ²South London & Maudsley NHS Foundation Trust; ³Norfolk & Suffolk NHS Foundation Trust

FS, https://orcid.org/0000-0003-4722-4549, LS, https://orcid.org/0000-0002-1643-9582; ER, https://orcid.org/0000-0001-6439-3438

Correspondence to: Dr Luke Skelton, ES1 PICU, Maudsley Hospital, South London & Maudsley NHS Foundation Trust, Denmark Hill, London, SE5 8AZ; luke.skelton@slam.nhs.uk

BRIEF REPORT

Challenges facing psychiatric intensive care during COVID-19

Luke Skelton, Lucy Blake, Margaret Butler, Ria Pugh, Bethan Harries, Faisil Sethi

Journal of Psychiatric Intensive Care Journal of Psychiatric Intensive Care, 16 (2): 95–99 doi:10.20299/jpi.2020.011 Received 25 June 2020 | Accepted 6 July 2020 © NAPICU 2020

BRIEF REPORT

Challenges facing psychiatric intensive care during COVID-19: mitigating the risk of transmission on a PICU

Ria Pugh, Luke Skelton, Lucy Blake, Margaret Butler, Bethan Harries, Faisil Sethi

Journal of Psychiatric Intensive Care Journal of Psychiatric Intensive Care, 16 (2): 101–109 doi:10.20299/jpi.2020.014 Received 23 July 2020 | Accepted 10 August 2020

BRIEF REPORT

Prescribing practices for psychiatric intensive care during COVID-19

Margaret Butler, Bethan Harries, Luke Skelton, Kalliopi Vallianatou, Lucy Blake, Ria Pugh, Faisil Sethi

BRIEF REPORT

Risk mitigation and the legal and ethical considerations for COVID-19 in a psychiatric inpatient setting

Lucy Blake¹, Ria Pugh¹, Luke Skelton¹, Eric Baskind², Bethan Harries¹, Margaret Butler¹, Faisil Sethi¹

Pandemic and the PICU: COVID-19 Clinical Model

Expeditious Approach to Testing & Screening for Symptoms

Least Restrictive
Interventions & Group
Space Interventions

Guiding Principles

Dynamic Systems
Approach (incl. rapid reviews, accelerated treatment plans & flow)

Wellbeing & Leadership Focus

VUCA







Uncertain



Complex



Ambiguous

AIM HIGH



TEAMWORKING



FAIL WELL & LEARN FAST



