



Dream Big : predicting Sepsis and building a healthcare AI framework for Queensland

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Acknowledgment

- Traditional Custodians
- Our vision is to use bigdata and AI research to help overcome inequity in health outcomes



Declaration

- Work funded by Qld Government
- *Enabled by the data of Queenslanders*

- Industry: Chief Medical Advisor for bdna, helped fund my attendance at this event





Qld Sepsis AI Working Group

- There's no i in TEAM ...
- Professor Ian Scott, Dr Anton Van Der Vegt, Rudolf Schnetler and Vikrant Kalke
- CBI eHealth – Cathi Ryan, Wil Wentholt
- Clinical Excellence Queensland – Michael Rice, Kirstine Sketcher-Baker
- Clinical Excellence Commission NSW
- Dr Amith Shetty, NSW Ministry Health

Sepsis

- Infection ...
- and *'body shuts down'*
- Organ dysfunction – need machine support in an ICU – *'life support'*
- Genetic response – inbuilt in your body
- Global Health Emergency – 1 in 5 deaths worldwide
- 5000-10000 deaths/year Australia
- There is NO single test for diagnosis
- <1% of emergency dept presentations – *needle in haystack*



MEDINFO23

8 - 12 JULY 2023 | SYDNEY, AUSTRALIA



WILL to LIVE



MATTHEW and DIANE AMES
with Kate Ames

It turns out Sepsis is hard to diagnose...

Just ask Matthew if it's time critical...

Every minute counts!



Why?

Impact of 1-hour and 3-hour sepsis time bundles on patient outcomes and antimicrobial use: A before and after cohort study

[Balasubramanian Venkatesh](#)   • [Luregn Schlapbach](#) • [Donna Mason](#) • [Kathryn Wilks](#) • [Robert Seaton](#) • [Paula Lister](#) • [Adam Irwin](#) • [Paul Lane](#) • [Lyndell Redpath](#) • [Kristen Gibbons](#) • [Endrias Ergetu](#) • [Michael Rice](#) • [Show less](#)

[Open Access](#) • Published: November 01, 2021 • DOI: <https://doi.org/10.1016/j.lanwpc.2021.100305>

- Improved time to antibiotics
- Decrease need for ICU
- Reduces mortality



A fancy Decision Tree ...

Queensland Government
Emergency Department Non-pregnant Adult Sepsis Pathway
 For tertiary and secondary facilities

(Affix identification label here)
 URN: _____
 Family name: _____
 Given name(s): _____
 Address: _____
 Date of birth: _____ Sex: M F

Facility: _____

Clinical pathways never replace clinical judgement. Cases outlined in this pathway must be altered if it is not clinically appropriate for the individual patient.

Septic Shock = shock + infection (mortality 20-23%) Sepsis = organ dysfunction + infection (mortality 10-12%)

Screen ALL non-pregnant adult emergency department patients who meet ANY of the following criteria (tick all that apply)

Looks sick Fever symptoms (or recent fever symptoms)
 You suspect they may have sepsis Hypothermia <35.5°C
 Has a suspected infection Signs of clinical deterioration (e.g. altered level of consciousness or total Q-AODS score of >4)
 If you suspect **metapneumoic sepsis**, refer to local guidelines if available, otherwise continue screening on this pathway

Screening initiated: / / (24hr)

Are ANY of the following risk factors present? (tick all that apply)
Absence of risk factors does not exclude sepsis as a cause of deterioration

Re-presentation within 48 hours Recent trauma or surgery / invasive procedure
 Malnourished or frail Postpartum / Miscarriage
 Immunosuppressed / Asplenia / Neutropenia IV drug use or alcoholism
 Involving medical device Aboriginal and / or Torres Strait Islander

AND / OR

Is there ANY reason to suspect an infection? (tick all possible sources that apply)

Yes, but source is unclear at present CNS / Meningitis
 Respiratory tract New onset confusion
 Urinary tract Family members / carers are concerned there is an infection
 Abdomen / GIT Other (specify): _____
 Skin / Joint / Prosthesis / Device

RECOGNISE

YES NO

Does the patient have ANY high risk criteria? (tick all that apply)

Respiratory rate ≥ 25 breaths/min
 New oxygen requirement to keep oxygen saturation $\geq 92\%$
 Heart rate ≥ 130 beats/min
 Systolic BP < 90 mmHg (or drop ≥ 40 from normal)
 Not passed urine in last 18 hours OR urinary output (UO) < 0.5 mL/kg/hr (if known)
 Evidence of new or altered mental state
 Lactate > 2 mmol/L if known
 Non-blanching rash / Mottled / Ashen / Cyanotic
 Recent chemotherapy

Does the patient have ANY moderate risk criteria? (tick all that apply)

Respiratory rate 21-24 breaths/min
 Heart rate 90-120 beats/min OR new dysrhythmia
 Systolic BP 90-99mmHg
 Not passed urine in last 12-18 hours
 Temperature $< 35.5^\circ\text{C}$ or $\geq 38.5^\circ\text{C}$
 Family members / carers concerned about mental state
 Acute deterioration in functional ability

ESCALATE / DE-ESCALATE

Patient has SEPSIS or SEPTIC SHOCK until proven otherwise

- Obtain immediate senior medical review
- Consider transfer to resuscitation area
- Commence resuscitation

Senior medical review attended: / / (24hr)

Does the senior medical reviewer think sepsis or septic shock is likely?

Sepsis / septic shock likely Sepsis / septic shock unlikely

YES **NO**

Commence resuscitation and treatment for sepsis NOW (See page 2)

Patient may have SEPSIS

- Ensure lactate taken
- Obtain senior medical review

Low risk for SEPSIS

- Look for other common causes of deterioration
- In the event of deterioration reassess sepsis risk using a new copy of this form
- If to be discharged home, give patient sepsis discharge instructions

Signature Log Every person documenting in this clinical pathway must supply a sample of their initials and signature below

Initials	Signature	Print name	Role	Initials	Signature	Print name	Role

ED NON-PREGNANT ADULT SEPSIS PATHWAY TERTIARY AND SECONDARY

DO NOT WRITE IN THIS BRUSHING MARGIN

V1.00 - 11/03/20
 WMC Code: 1N07T04
 20/06/20



The Problem





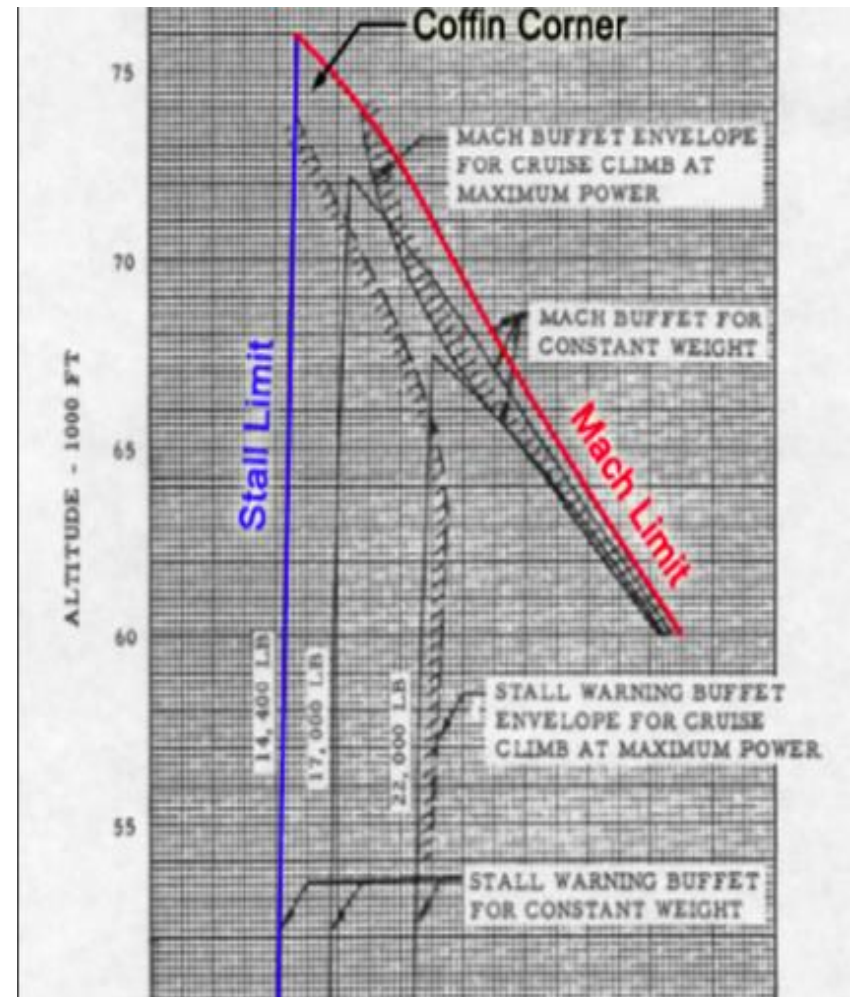
Why is Aviation so SAFE?? *“today I’ll turn off the onboard augmented intelligence...”*





Coffin Corner... *The Risk of Unstable Flight*

- At high altitude the stall speed (plane falls from sky) is very near to critical Mach number (maximum speed) where lift is lost due to shock waves and drag is increased
- Both followed by structural failure as airplane falls
- Small margin for error

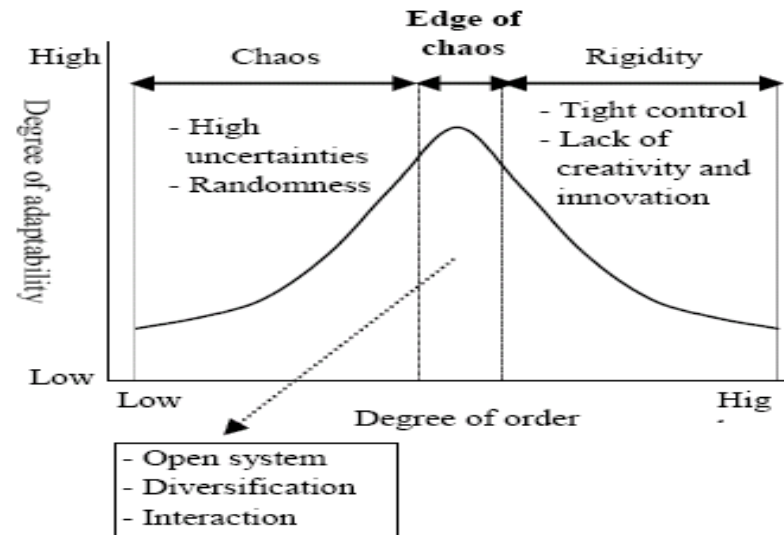




The 'Edge of Chaos' Problem

- Clinicians working in the complexity of modern healthcare live on the edge of chaos where demand always exceeds capacity.*

The Relations between Order and Adaptability



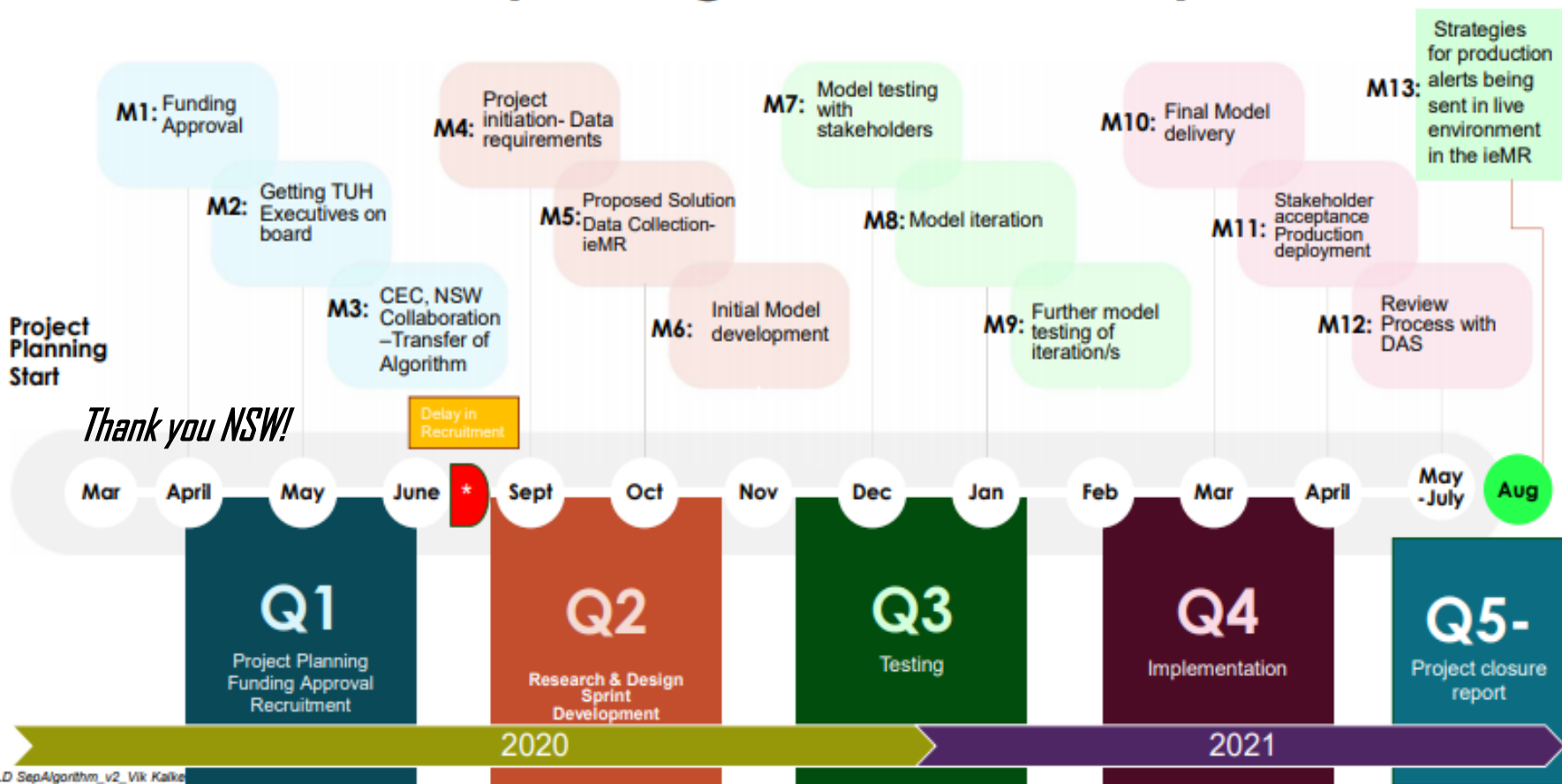


How do we support busy clinical teams?

- We build and test Adult Sepsis predictive algorithms using Artificial Intelligence to provide timely clinician support to assist with the diagnosis of Sepsis
- *Paul, proceed until apprehended* Prof Keith McNeil

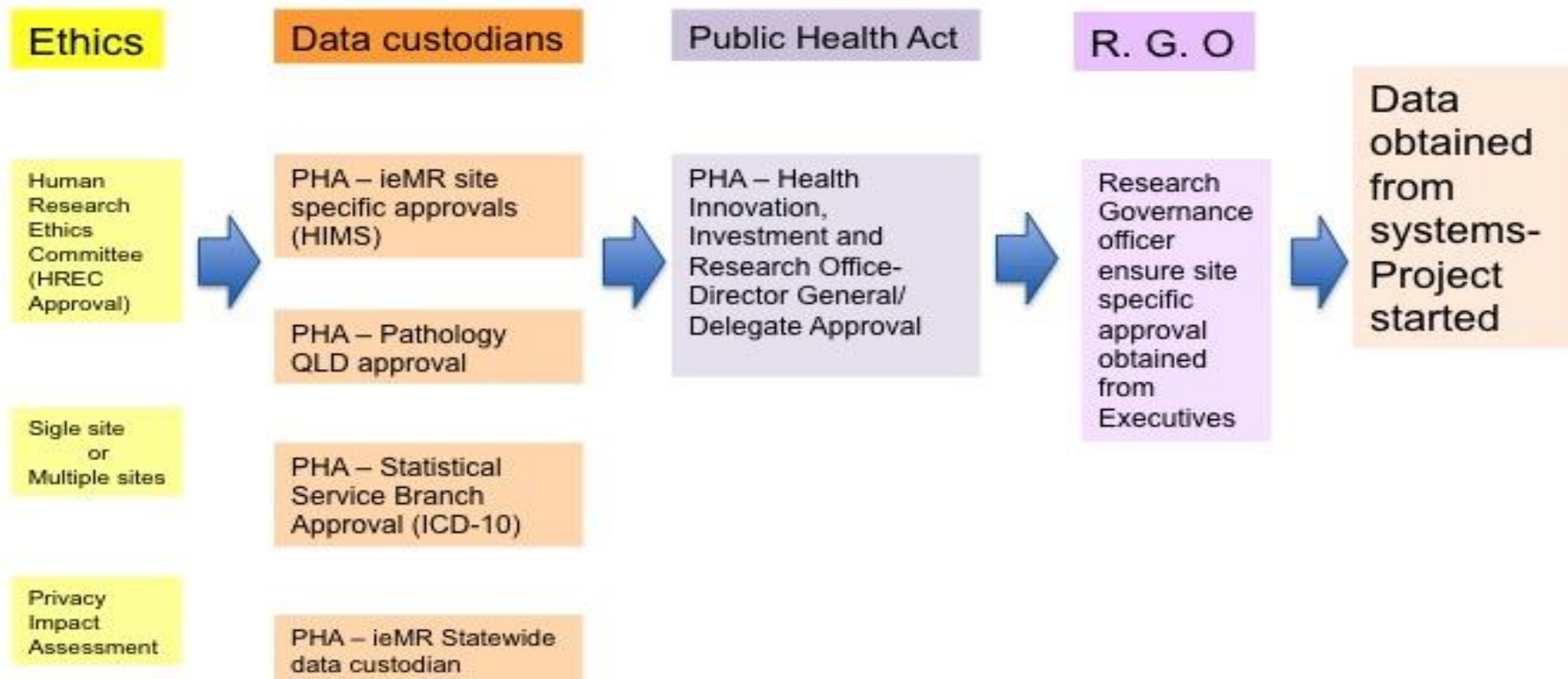


QLD Sepsis Algorithm Roadmap





Low risk research project – Journey (As Imagined)

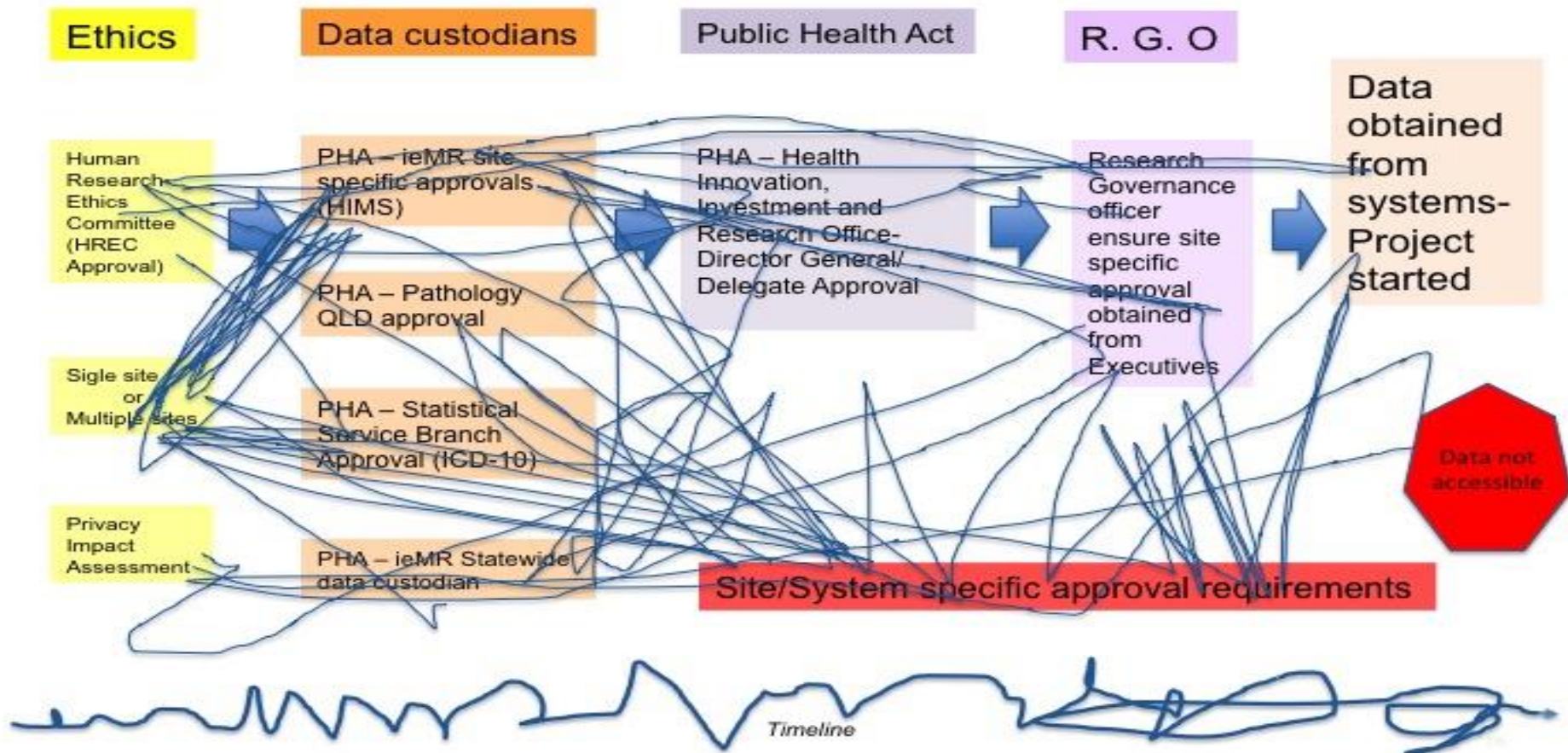


Timeline





Low risk research – Journey (As Done)





12 Digital Hospitals – **REAL EMR Data**

Merged 4 yrs EMR data: 524,000 patients with **1.13M** encounters

10,922 encounters coded sepsis (ICD-10)

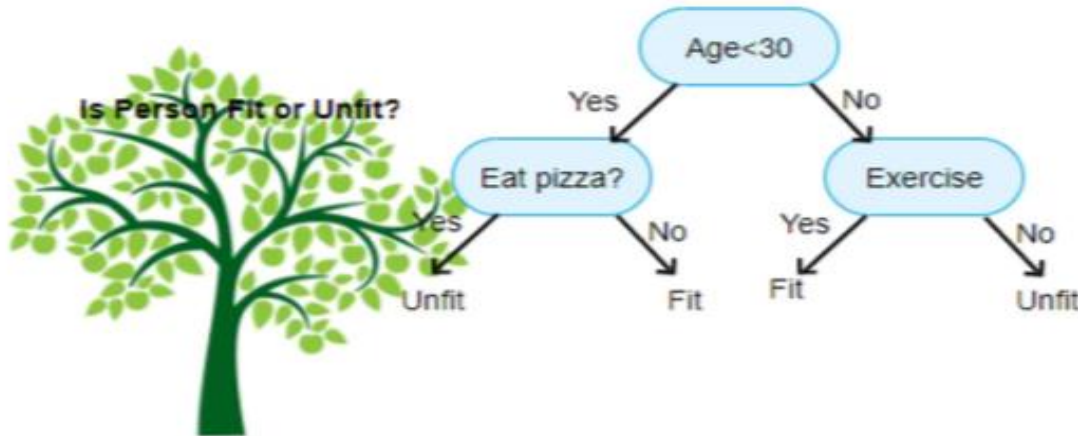
26,753 sepsis encounters by an alternative Sepsis 3 classification

3.4 trillion bytes

Median age 64, mortality of 10%



Microsoft
LightGBM



```
print('Create RepeatedStratifiedKFold splits...')
rskf = RepeatedStratifiedKFold(n_splits=10, n_repeats=3, random_state=42)

sen_list = []
spec_list = []
ppv_list = []
npv_list = []
auroc_list = []
acc_list = []
interval_list=[]
nsw_feature_importance = pd.DataFrame()

print('Start training across splits...')
for train_index, test_index in rskf.split(X, y):
    print("TRAIN:", train_index, "TEST:", test_index)
    X_train, X_test = X[train_index], X[test_index]
    y_train, y_test = y[train_index], y[test_index]

    print('Train Model...')
    train_data = lgb.Dataset(X_train, label=y_train)
    max_depth = 120
    param = {'max_depth':max_depth, 'num_leaves': (2^max_depth), 'objective': 'binary',
             'num_round': 500}
    model = lgb.train(param, train_data, num_round)
    print('Predicting over test...')
    y_pred = model.predict(X_test).round(0)
    y_pred=y_pred.astype(int)
    nsw_feature_importance = nsw_feature_importance.append(pd.DataFrame({'Gain':model.feature_importance(importance_type='gain')}))
    sen_list.append(results(y_pred,y_test)['Sensitivity'])
    spec_list.append(results(y_pred,y_test)['Specificity'])
    ppv_list.append(results(y_pred,y_test)['PPV'])
    npv_list.append(results(y_pred,y_test)['NPV'])
    auroc_list.append(results(y_pred,y_test)['AUROC'])
    acc_list.append(accuracy_score(y_test,y_pred))
    acc = accuracy_score(y_test,y_pred)
    inter = z * sqrt((acc * (1-acc))/len(X_train))
    interval_list.append(inter)

print("Saving lightgmb data")
result = pd.DataFrame(data={'Algorithm':f'NSW LightGBM {h}',
                           'Sensitivity': [np.mean(sen_list)],
                           'Specificity': [np.mean(spec_list)],
                           'PPV': [np.mean(ppv_list)],
                           'NPV': [np.mean(npv_list)],
                           'AUROC': [np.mean(auroc_list)],
                           'Accuracy': [np.mean(acc_list)],
                           'Interval': [np.mean(interval_list)]
                          })
```

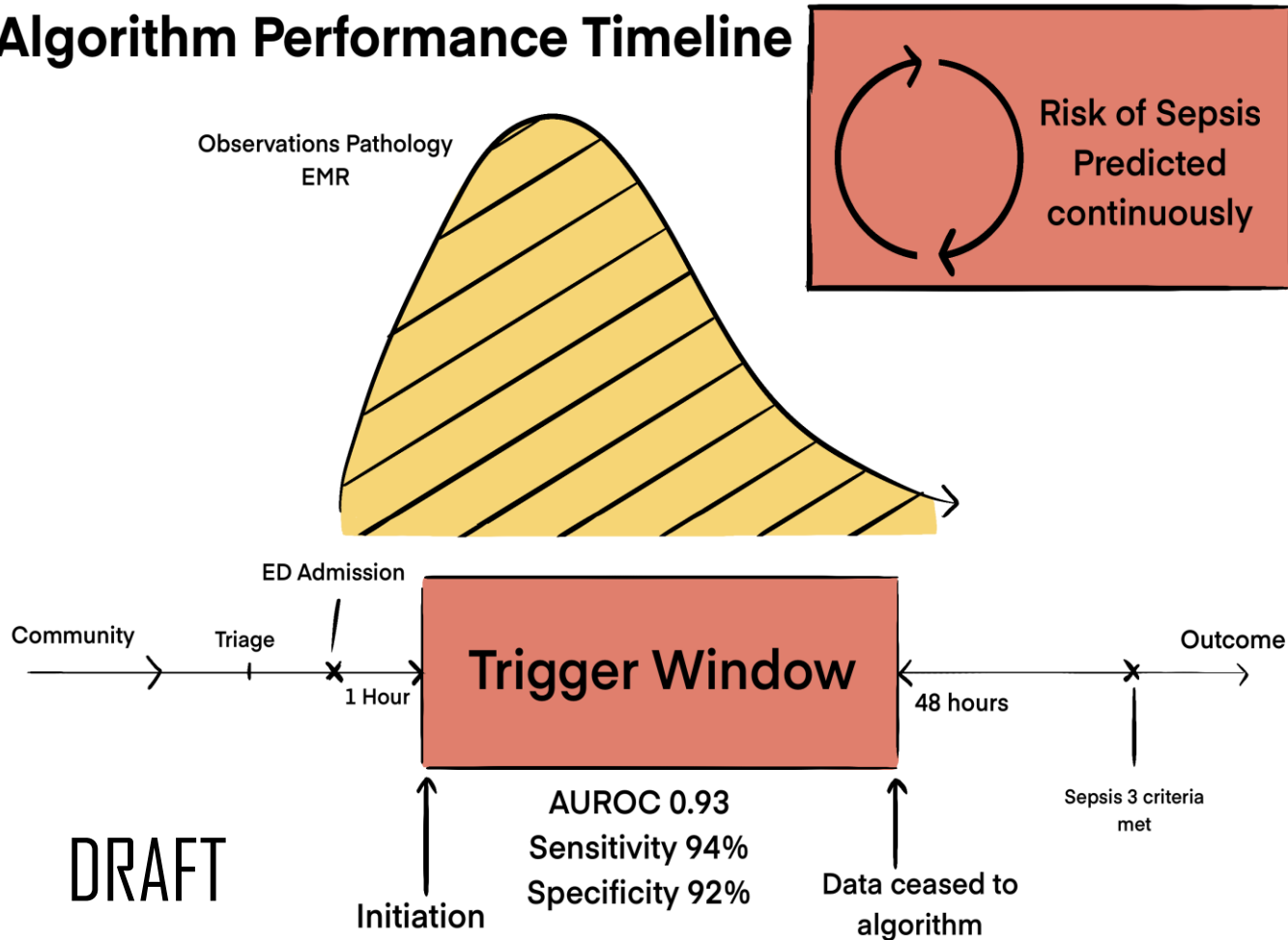
Machine Learning - lots of decision trees

The Algorithm learnt from the 'weakest' trees!

Explainable AI...



Algorithm Performance Timeline



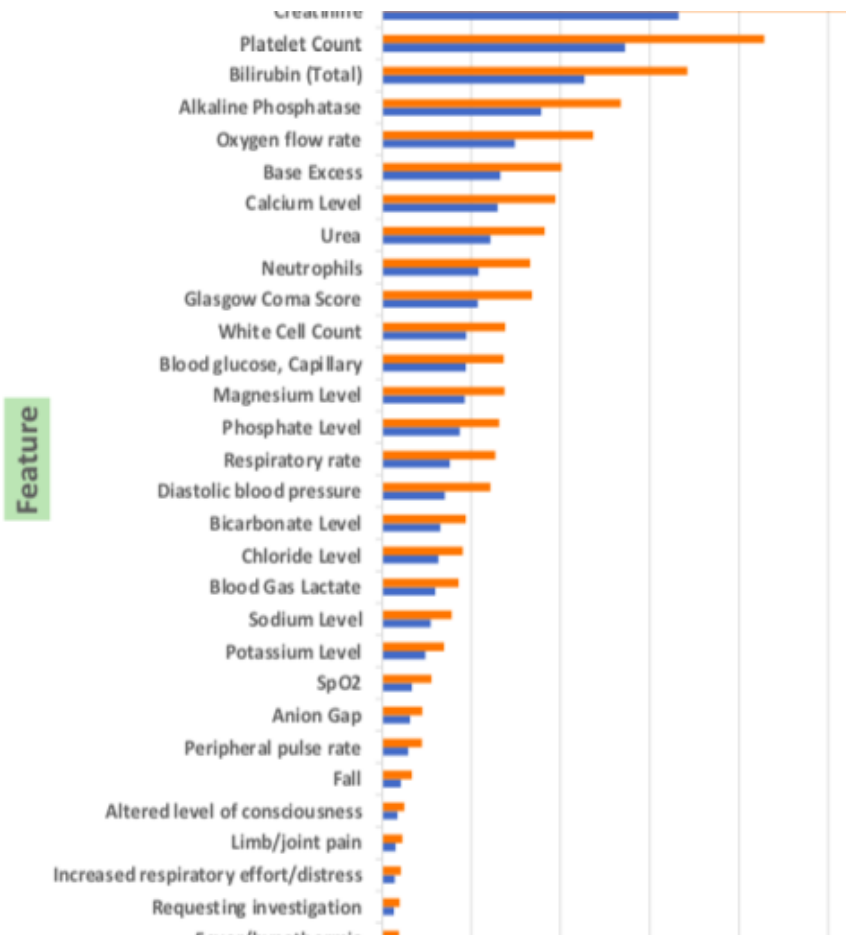


Algorithm	Sensitivity	Specificity	PPV	NPV	AUROC
QLD LightGBM 48	93.96	91.69	53.24	99.34	0.93
QLD Weighted LightGBM 48	80.23	94.03	68.10	96.77	0.87

Table 11 Machine learning Results with Additional Features

DRAFT

Weighted - removes bias of small sample

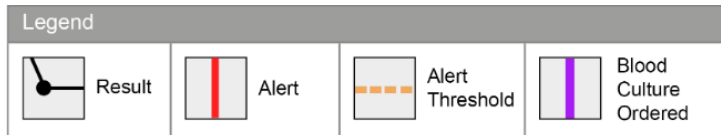
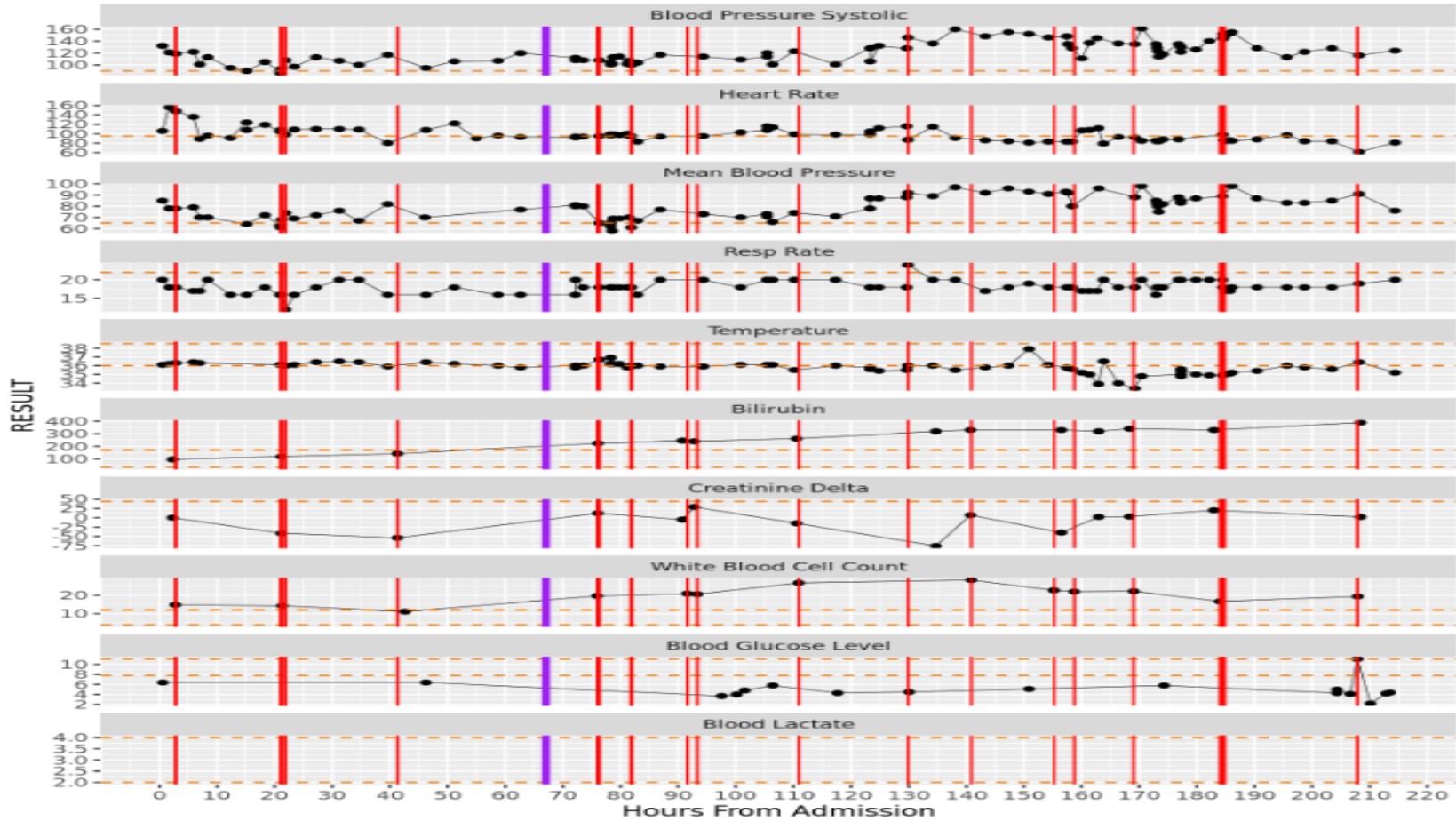


124 Features

...vs human
Cognitive load...



Results/Alerts by hours from admission





The road to hell is paved with good intentions ...

JAMA Internal Medicine | [Original Investigation](#)

External Validation of a Widely Implemented Proprietary Sepsis Prediction Model in Hospitalized Patients

Andrew Wong, MD; Erkin Otles, MEng; John P. Donnelly, PhD; Andrew Krumm, PhD; Jeffrey McCullough, PhD; Olivia DeTroyer-Cooley, BSE; Justin Pestrue, MEcon; Marie Phillips, BA; Judy Konye, MSN, RN; Carleen Penosa, MHSA, RN; Muhammad Ghous, MBBS; Karandeep Singh, MD, MMSc



The need to overcome the bad stuff ...

- Community suspicion, lack of trust
- Inequity
- Bias
- Consumer Privacy Fears
- Ethical Debt
- Fear of the unknown
- *Doctors are a suspicious bunch generally...*



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<https://doi.org/10.1093/jamia/ocad088>


Research and Applications



OXFORD

Research and Applications

Implementation frameworks for end-to-end clinical AI: derivation of the SALIENT framework

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Staged Clinical AI Implementation Framework - SALIENT

- Full disclosure, honesty and transparency, rigor in evaluation
- Clinician lead, not for profit


The Elephant in the room ...

- 15 000 AI papers in healthcare since 2020
- 45 published implementations
- *End to end implementation framework allowing for staged evaluation of the AI algorithms*
- Derived from various accepted AI research models – TRIPOD, DECIDE AI, CONSORT AI



Review

Deployment of machine learning algorithms to predict sepsis: systematic review and application of the SALIENT clinical AI implementation framework

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¹Queensland Digital Health Centre, The University of Queensland, Brisbane, Queensland, Australia, ²Department of Internal Medicine and Clinical Epidemiology, Princess Alexandra Hospital, Brisbane, Australia, ³Centre for Information Resilience, The University of Queensland, St Lucia, Australia, ⁴School of Information Technology and Electrical Engineering, The University of Queensland, St Lucia, Australia, ⁵Patient Safety and Quality, Clinical Excellence Queensland, Queensland Health, Brisbane, Australia, ⁶Safety Quality & Innovation, The Prince Charles Hospital, Queensland Health, Brisbane, Australia

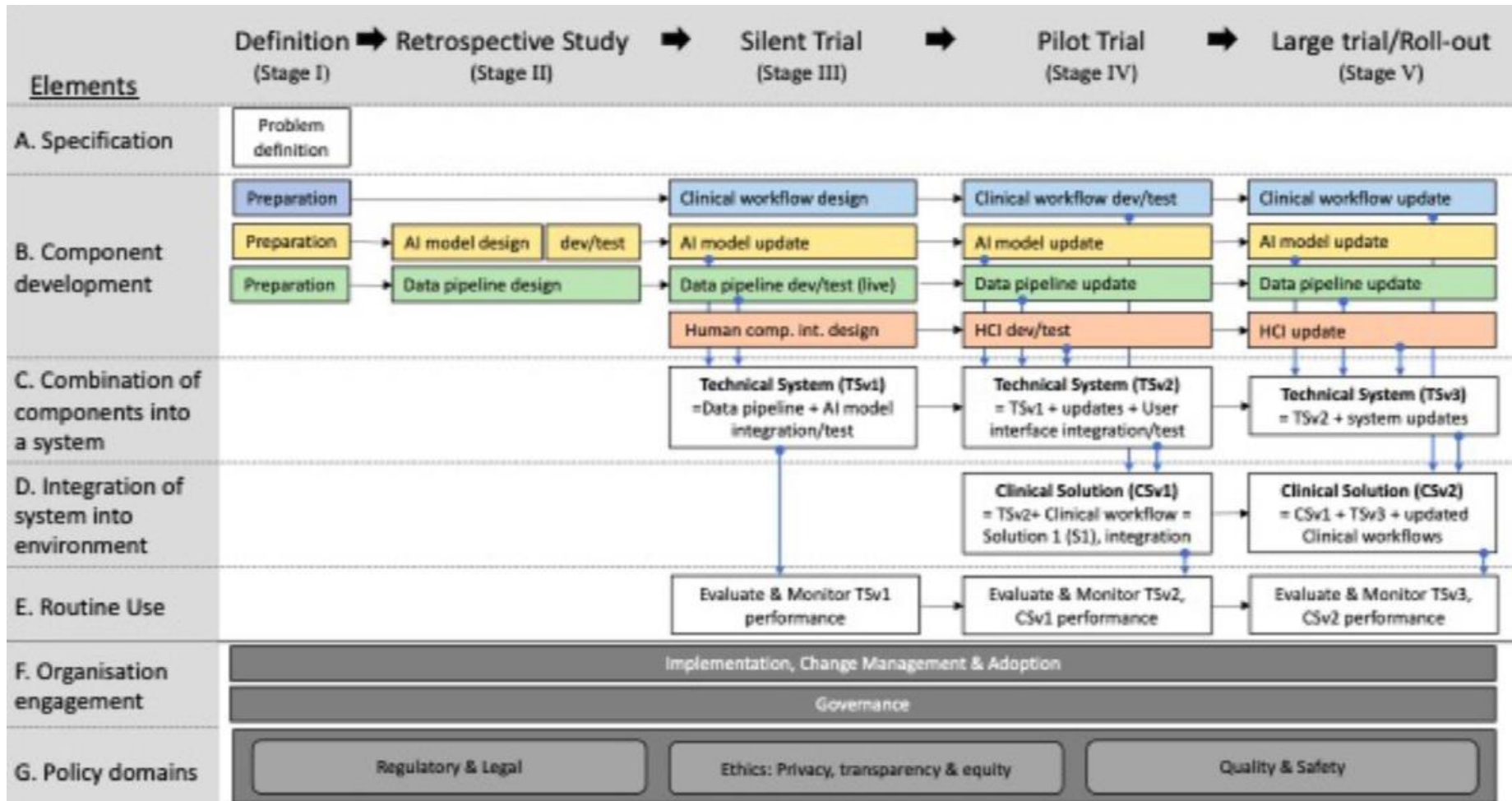
Corresponding Author: Anton H. van der Vegt, PhD, BE, BSc, Queensland Digital Health Centre, The University of Queensland, Princess Alexandra Hospital, 34 Cornwall St, Woolloongabba, Brisbane, QLD 4072, Australia; a.vandervegt@uq.edu.au

How does it perform?

Adding Rigor to the process...

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Silent Trial – Stage III

- Prospective Evaluation
- 5 large Qld Digital Hospitals
- Semi real-time data feed from Clinical Business Intelligence, Ehealth QLD
- Begin Aug-Sept 2023

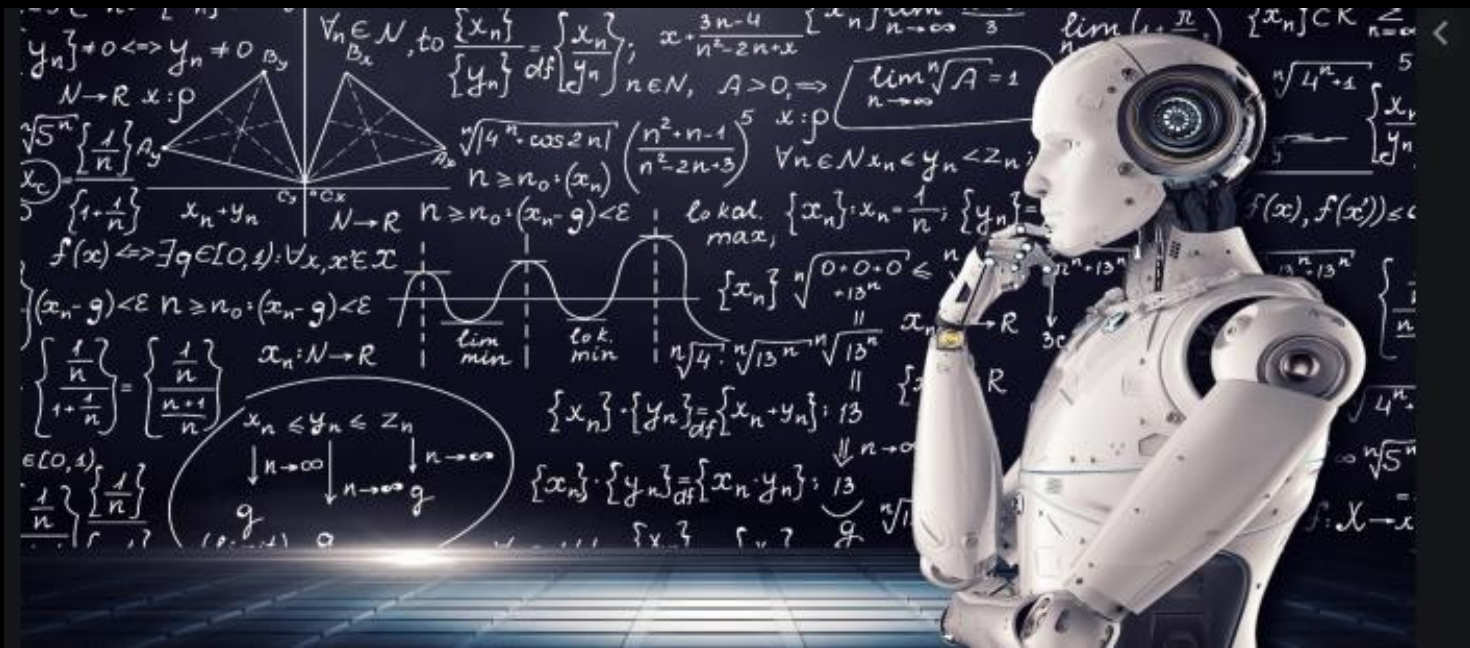


Reflections

- Data governance is hard - a major blocker in Qld!
- De-identified real data will be essential for researchers/universities
- Once evaluated 'in-silico', presented to real time data for further evaluation – a staged approach, measured evaluation and rigor
- Building an end-to-end framework for the future
- Ethical Debt – the need to engage the community
- *Deteriorating Patient* work next... our false positives!
- First Nations QSA – burden of sepsis

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QUESTIONS?