

# The development of a machine learning-driven clinical decision support system for the management of acute dengue in Vietnam – ASTMH 2022

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## Introduction

Seasonal dengue epidemics exert significant pressure on healthcare settings worldwide

Adoption of appropriate clinical decision-support systems (CDSS) at patient point-of-care could improve the quality and consistency of care.

As a multidisciplinary of clinicians, engineers, software designers, we developed the D-CAT (Dengue Clinical Assessment Tool) with the aim of augmenting decision-making in a hospital setting within Vietnam.

## Methods

We performed process mapping and semi-structured interviews to understand better the determinants in decision-making in dengue.

In parallel, machine learning models (XGBoost and Neural network) machine learning models were trained from clinical data from 4,131 patients between 1999-2018 for prediction of dengue shock.

Interface design was done through a human-centred design process with a UX designer.

## The system

The system uses a lightweight reactive UI and is accessible through computers, tablets and mobile phones. Individual user profiles are password-protected. Language switching between English and Vietnamese can be toggled. Additional functions and modules can be added on to the architecture in future.

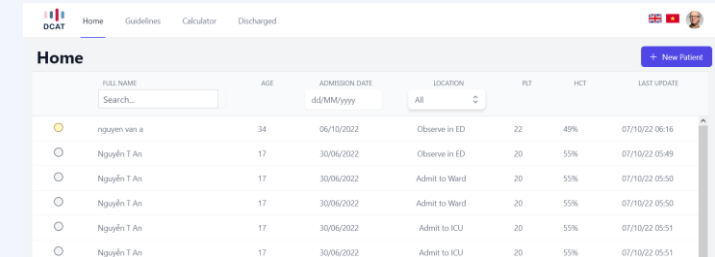
A patient information system provides at-a-glance data on recent results, observations and location, with secure data storage on the intranet.

The machine learning module utilises age, sex, day of illness, haematocrit, platelet data over admission to predict risk of shock (developmental hold out set testing AUROC = 0.82). Output currently uses a traffic light system to indicate predicted risk.

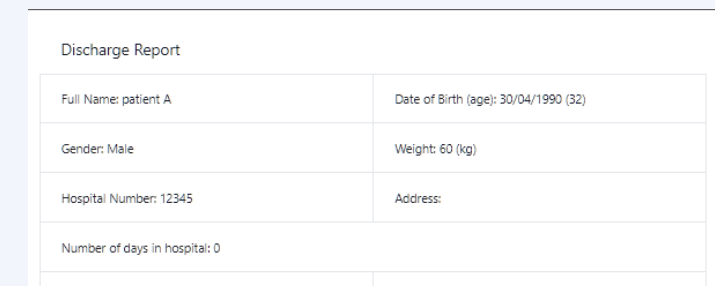
A discharge report which contains all laboratory results and clinical events

## Next steps

Prospective evaluation is planned 2022-2023 through workshops, usability testing in clinical areas and evaluation of model performance.



FULL NAME	AGE	ADMISSION DATE	LOCATION	PLT	HCT	LAST UPDATE
nguyen van a	34	06/10/2022	Observe in ED	22	49%	07/10/22 06:16
Nguyễn T An	17	30/06/2022	Observe in ED	20	55%	07/10/22 05:49
Nguyễn T An	17	30/06/2022	Admit to Ward	20	55%	07/10/22 05:50
Nguyễn T An	17	30/06/2022	Admit to Ward	20	55%	07/10/22 05:50
Nguyễn T An	17	30/06/2022	Admit to ICU	20	55%	07/10/22 05:51
Nguyễn T An	17	30/06/2022	Admit to ICU	20	55%	07/10/22 05:51



Discharge Report	
Full Name: patient A	Date of Birth (age): 30/04/1990 (32)
Gender: Male	Weight: 60 (kg)
Hospital Number: 12345	Address:
Number of days in hospital: 0	

## References

Applied machine learning for the risk-stratification and clinical decision support of hospitalised patients with dengue in Vietnam, Ming et al., 2022 PLOS Digital Health  
Mapping patient pathways and understanding clinical decision-making in dengue management to inform the development of digital health tools Huy et al., 2022 submitted

