Reducing Outpatient MRI Appointment No-show Rate Through Artificial Intelligence Predictive Analytics

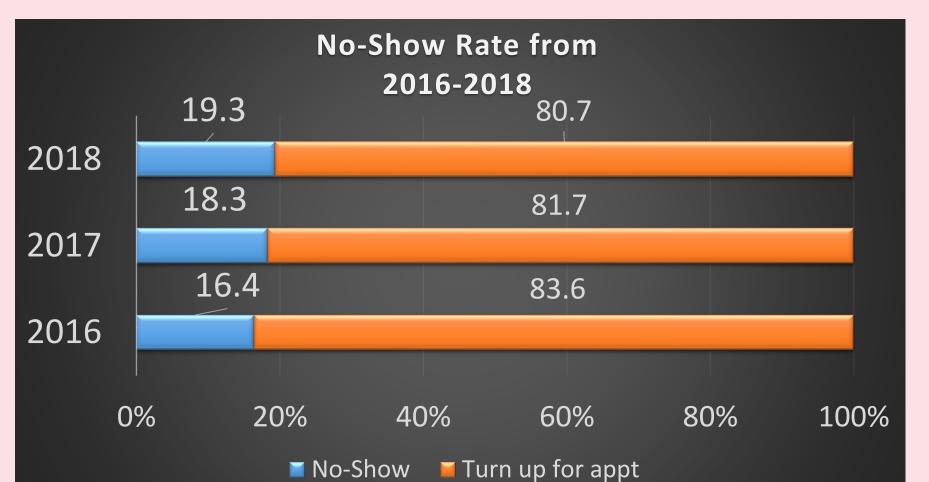


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Problem Statement / Background:

□ Problem Identification: Outpatient MRI no-show rates have shown an increasing trend between 2016 to 2018 (*Figure 1*) leading to issues detailed in *Figure 2*.



- Retrospective data of 32,957
 MRI outpatient appointments
 between 2016-2018 were
 extracted from RIS and OAS.
- ☐ Total no-shows of <u>5734</u>
 (17.4%) with yearly increasing trend as shown in *Figure 1*.

Figure 1. Increasing Trend in No-show Rates

PROJECT OBJECTIVES:

- ☑To reduce no-show rate
- ☑Reduce MRI appointment lead time
- ☑Improve accessibility of MRI services
- ✓ Improve patient care / pathway

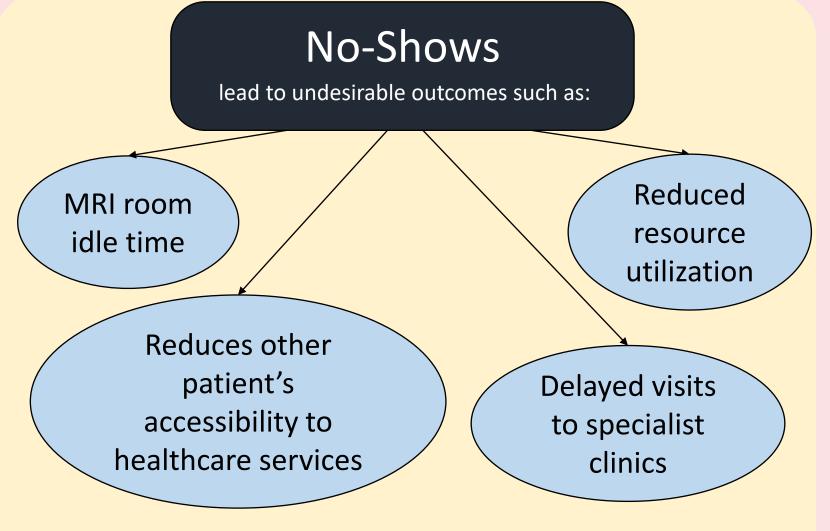
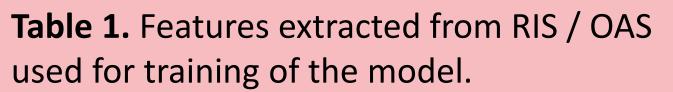


Figure 2. Undesirable Outcomes From Increasing No-show Rates.

Problem Analysis & Evaluation:

☐ Causes of appointment no-shows are complex, with multiple human and non-human factors potentially interacting in complex ways to influence appointment attendance.

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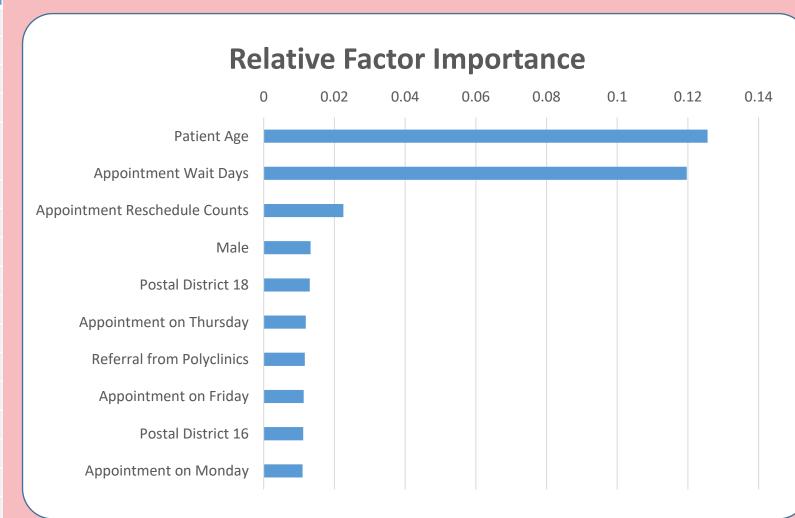


Figure 3. Relative importance of various features in predicting appointment no-show identified by the XGBoost model.

- ☐ A machine learning model trained with an extreme gradient boosting algorithm (XGBoost) was developed to predict outpatient MRI appointment no-shows.
- ☐ Table 1 shows a list of features extracted from RIS and OAS used for model training.
- □ Figure 3 shows the relative importance of these features as identified by the XGBoost machine learning model in predicting no-shows.
- ☐ The ROC-AUC of the model was 0.746, sensitivity of 85.2%, positive predictive value of 60.6% and overall accuracy of 65.4%.

Evaluation of solutioning involves consideration on the following aspects:

Sustainability

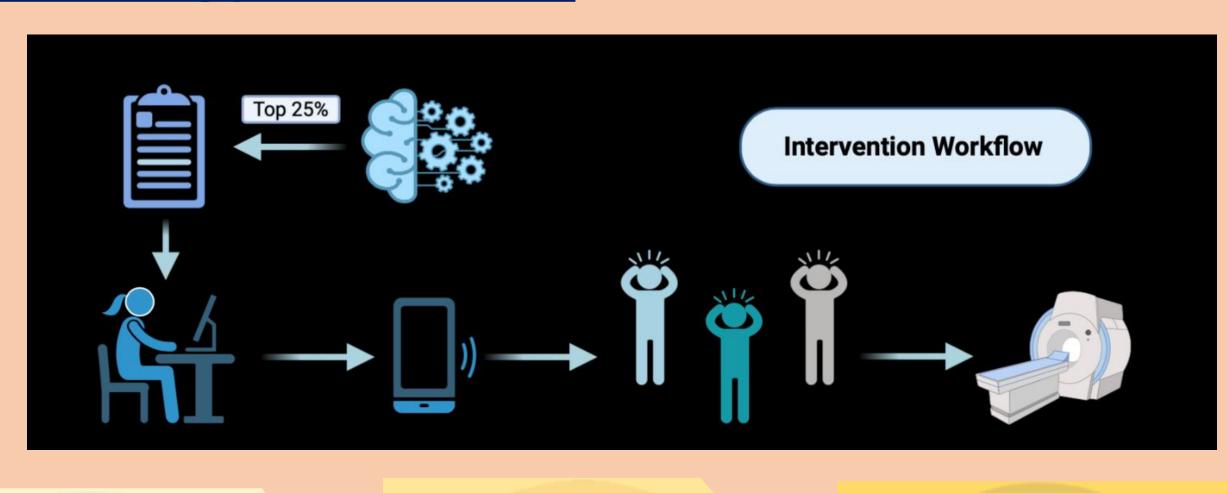
Manpower

Cost Effective

Time Effective

Main advantage of using AI predictive model → Computer generated list of patients with high probability of no-shows (Data supported selection)

Methodology & Results:



- Machine learning model (XGBoost) developed with features extracted from RIS and OAS from 2016-2018 appointments.
- Weekly list of outpatient MRI appointments with highest risks of noshow predicted by model generated.
- MRI Radiographers call top 25% of patients at highest risk of no-show one working day before appointments to confirm attendance.
- ☐ There was a significant absolute decrease in no-show rates of 3.4% over 6 months following implementation, with a relative decrease of 17.2% compared to the 12 months pre-intervention baseline no-show rate (p < 0.0001, Figure 4).
- □ Appointment no-show rates for contactable and non-contactable patients at high-risk of no-shows were 17.5% and 40.3% respectively (p < 0.0001).

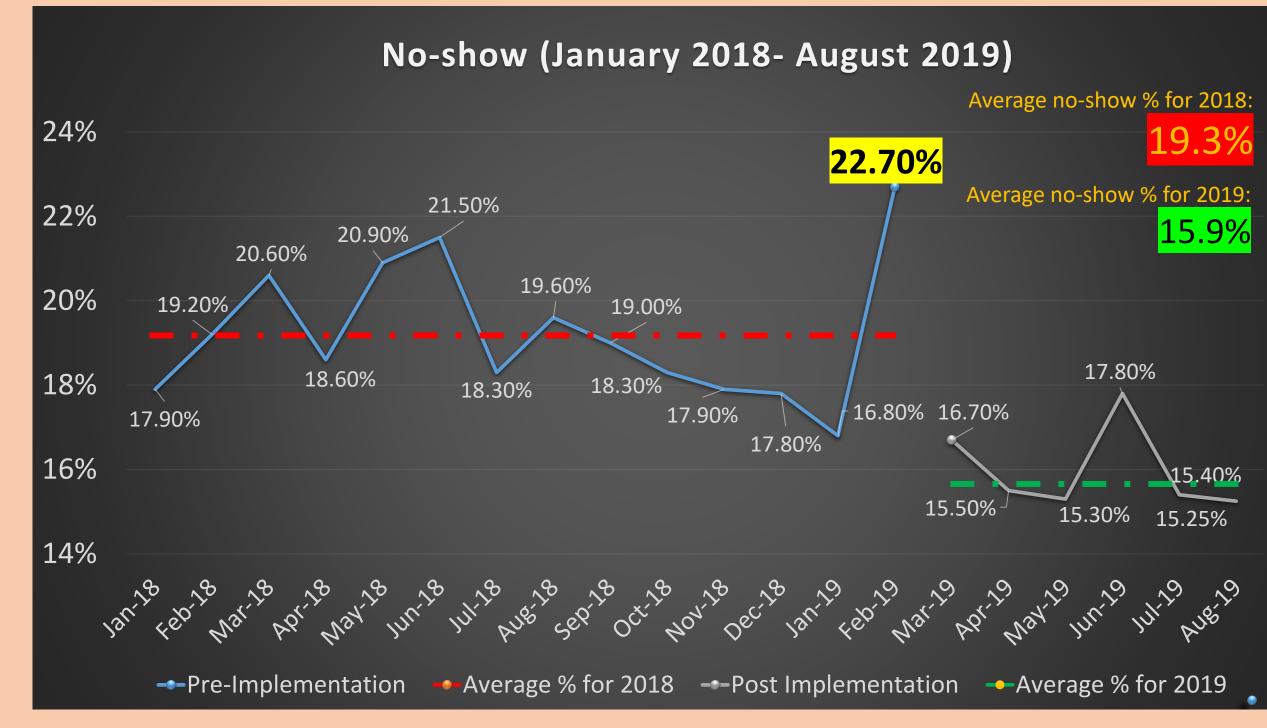


Figure 4. No-show rates 12-months preceding and 6 months after intervention measures.

Positive Outcomes & Conclusion:

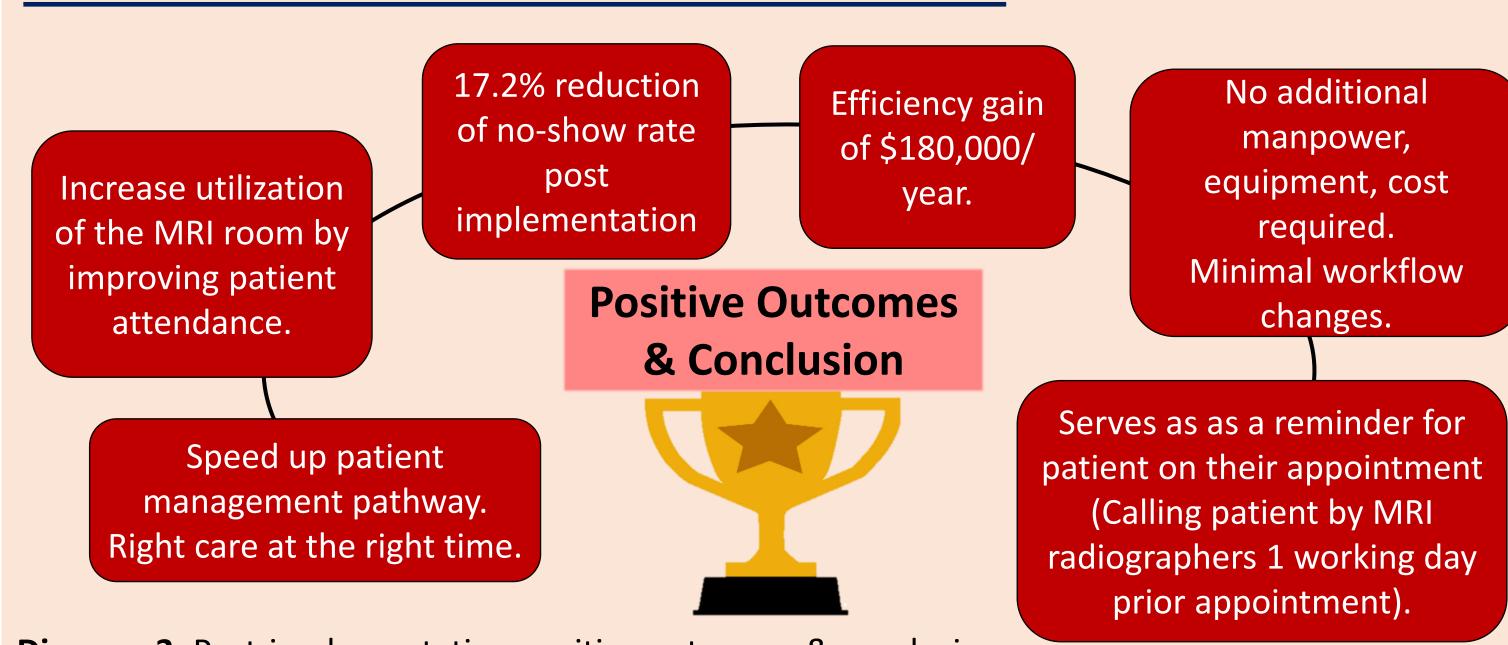


Diagram 3. Post-implementation positive outcomes & conclusion.

Subsequent Implementation & Improvement:

- The project objectives are met and post-intervention measures producing promising results. The department has since the onset of the COVID-19 pandemic adopted to call **all MRI outpatients** instead of the top 25% of patients generated from the predictive model.
- ➤ By calling all MRI outpatients, the monthly tabulated data (2020-early 2021) showed **further decreased no-show rates ranging between 8%-13%.** Nevertheless, the data may be affected by COVID-19 surge and measures.
- In the future, this predictive model can be readily adaptable to other diagnostic imaging modalities such as CT or US to manage patient scan appointment attendance.