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Prof Jacomine Grobler~ Department of Industrial Engineering

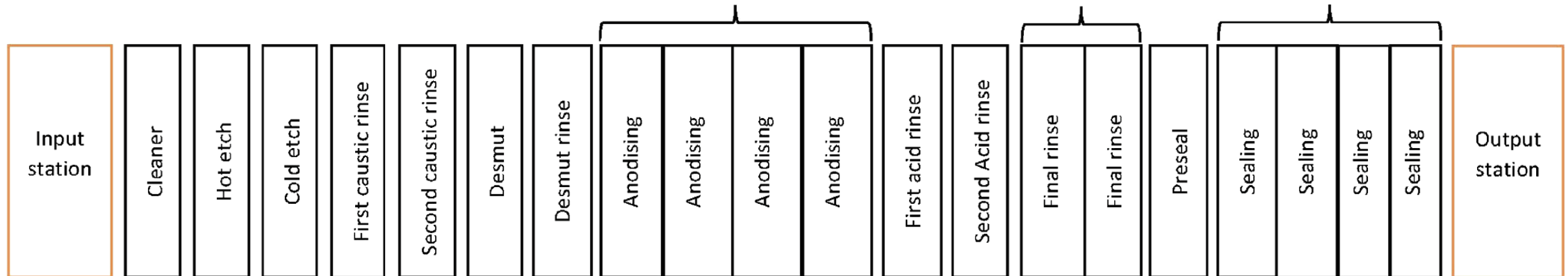
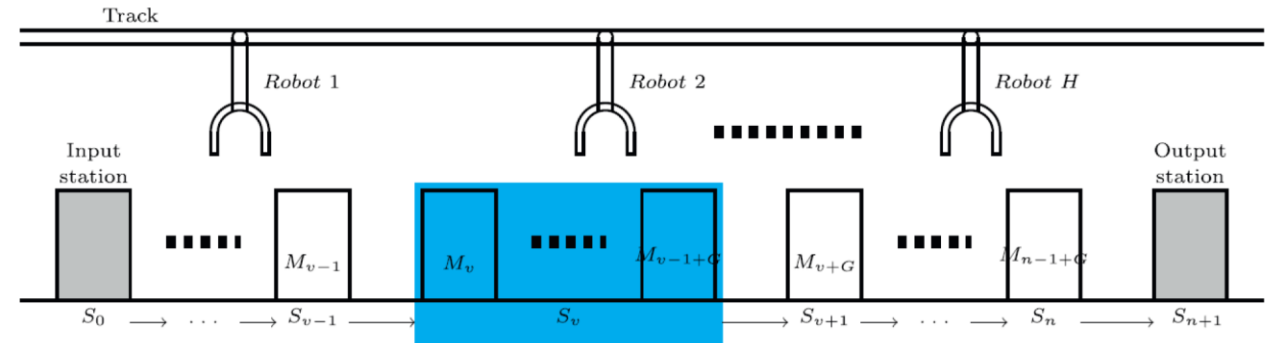
AI-based industrial engineering: using data for improved decision making

Industry Showcase 2025



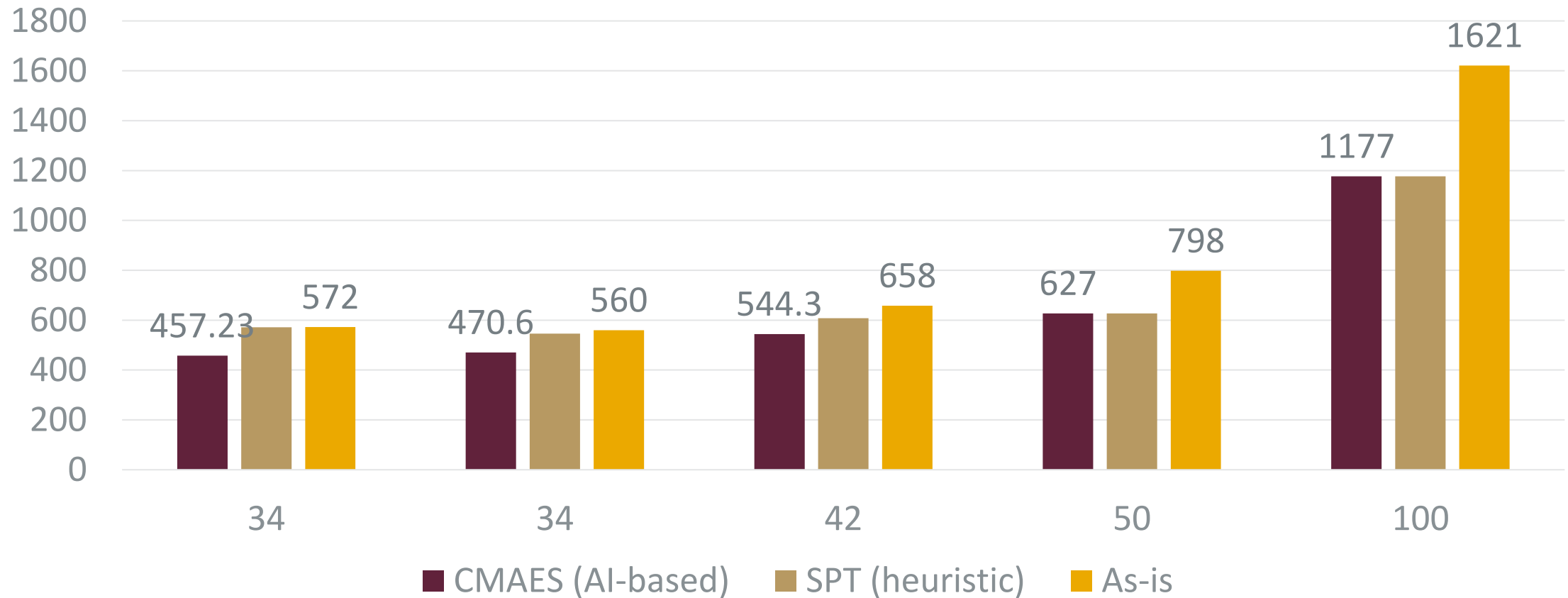
DATA

Aluminium anodising scheduling



Aluminium anodising scheduling

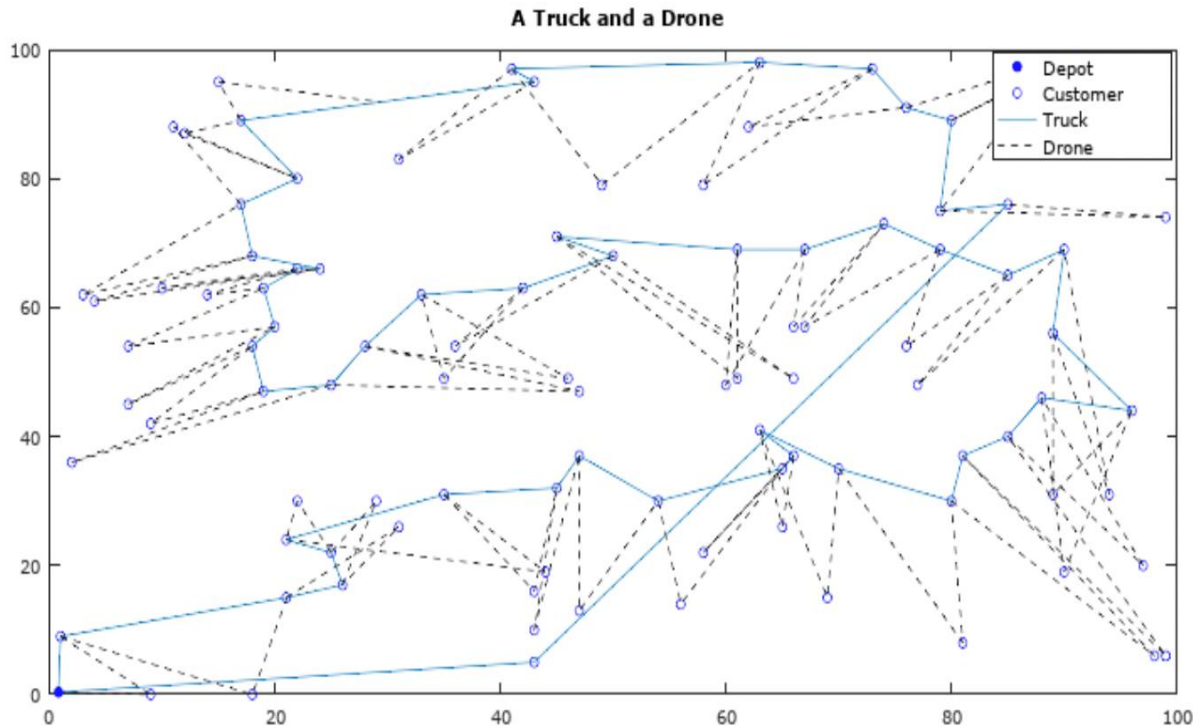
Completion time vs # customer nodes



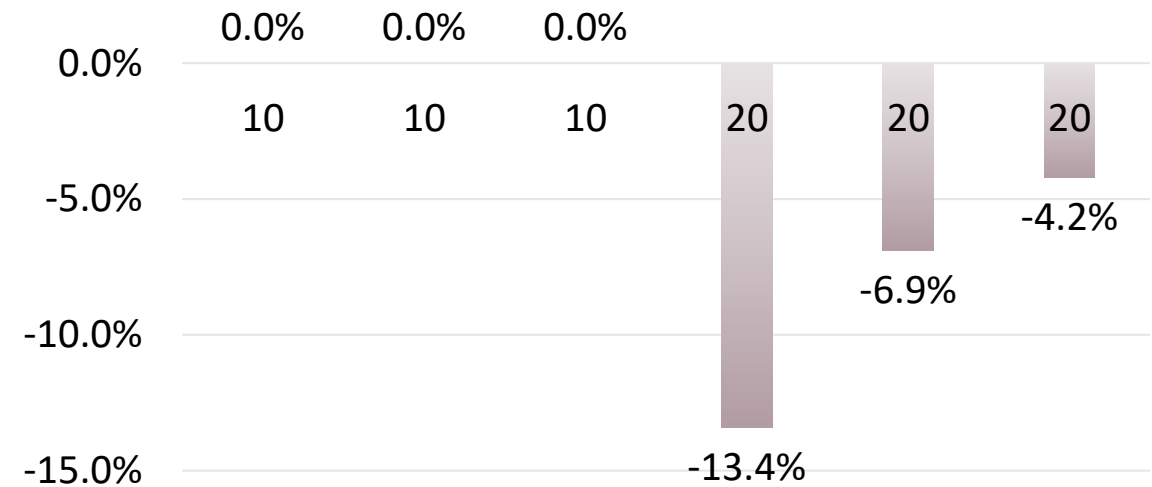


amazon

Optimal vs AI-based Algorithm Results

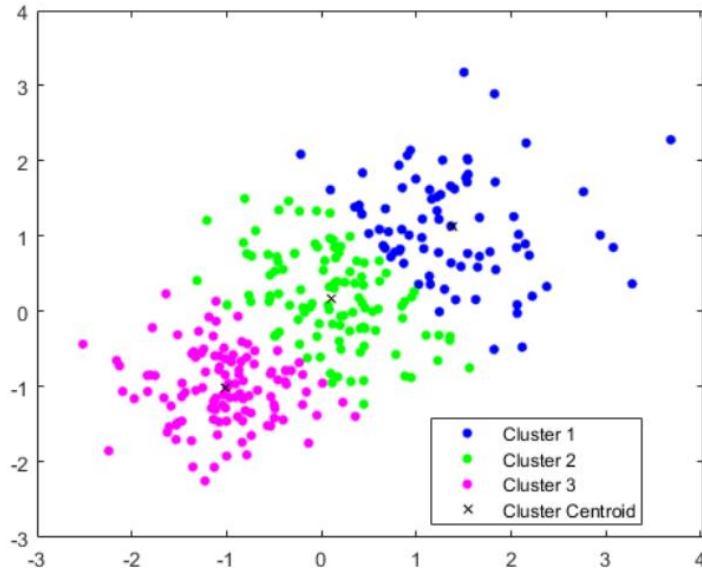


% decrease in performance per customer nodes

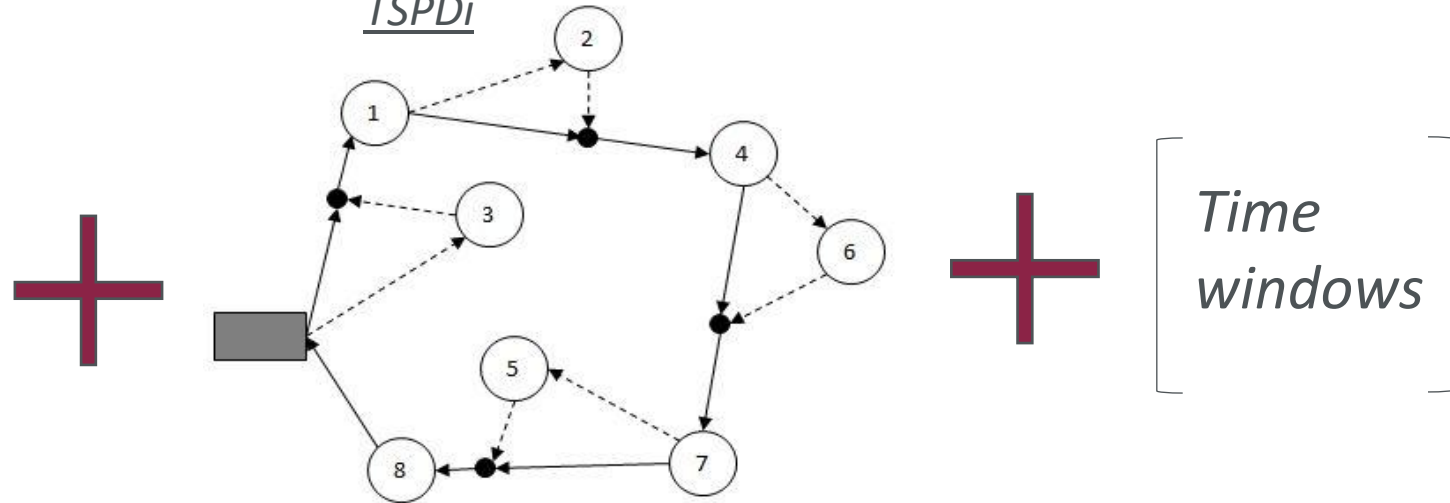


VRPDi with time windows

K-means Clustering

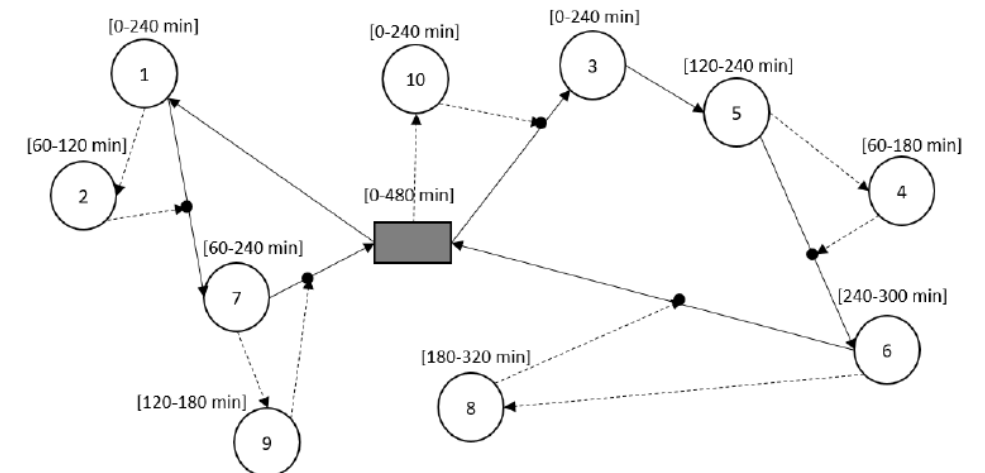


TSPDi



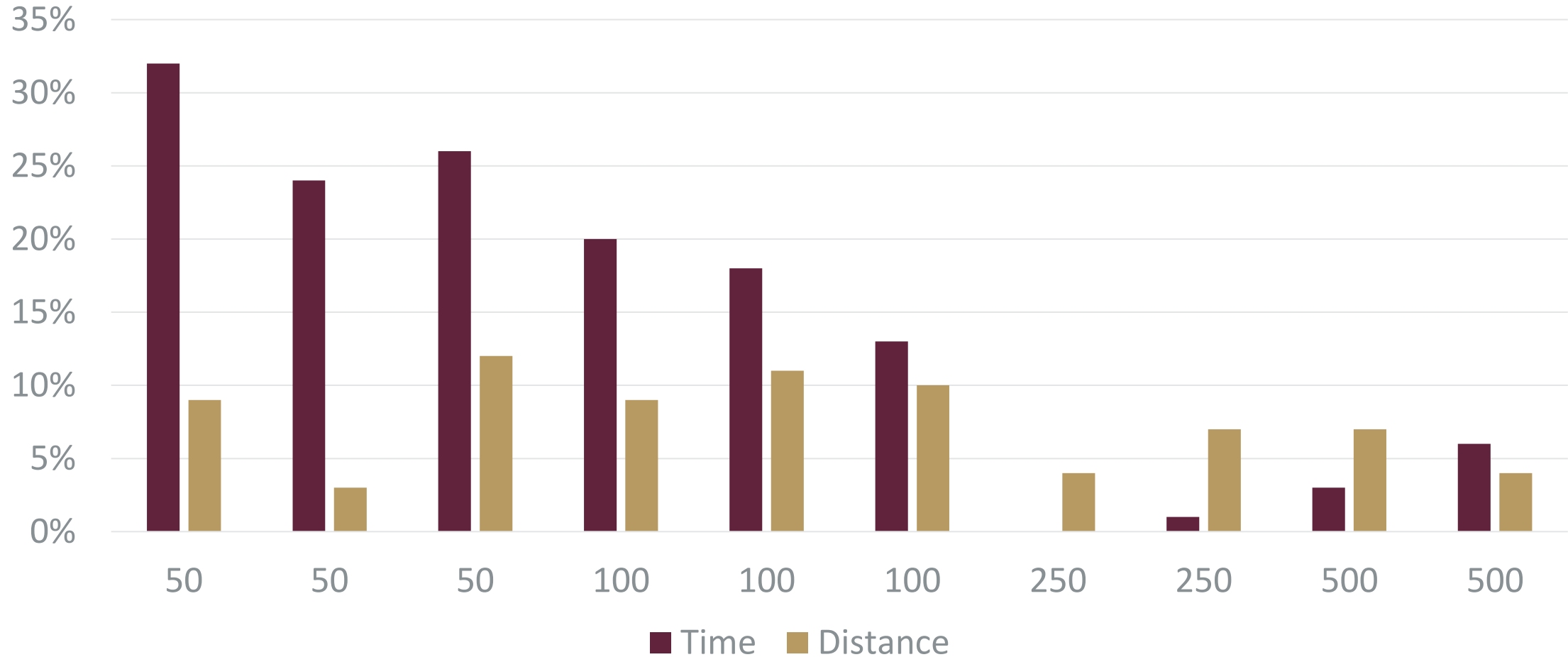
Time windows

VRPDiTW



Truck-only vs truck-and-drone system (with time windows)

% improvement vs # of customer nodes





Mozambican Cassava Suppliers



Cassava stiff porridge



Boiled cassava roots



Cassava leaves

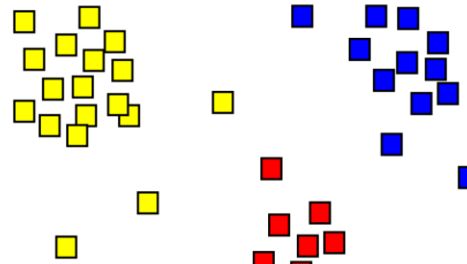
Segmentation of Mozambican Cassava Suppliers



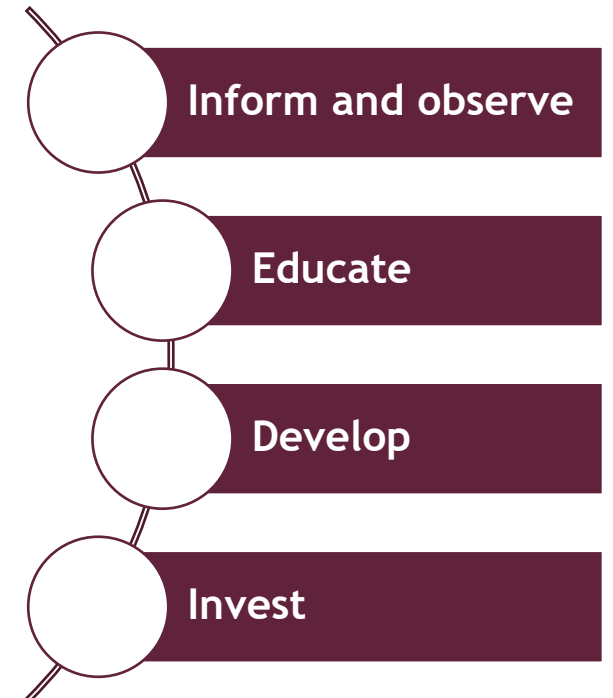
Feature name
Farmer code
Location of factory
Location of plot
Latitude of plot
Longitude of plot
Fieldworker
Modified variety?
Starch content (%)
Cassava quantity (Kg)
Cassava cost (MZN)
Transport cost (MZN)



- K-means
- AHC
- SOM



- Supply risk
- Effectiveness of operations
- Performance improvements



3000 suppliers over 2 years

Deep learning for identifying paediatric tuberculosis

- Tuberculosis diagnosis in young children is difficult.
- Most prevalent in LMIC.
- Common symptom of lymphobronchial tuberculosis is enlarged lymph nodes which lead to airway compression.
- This study investigated the use of deep learning for identifying lymphobronchial tuberculosis in children under the age of 5 with chest X-ray images as input.

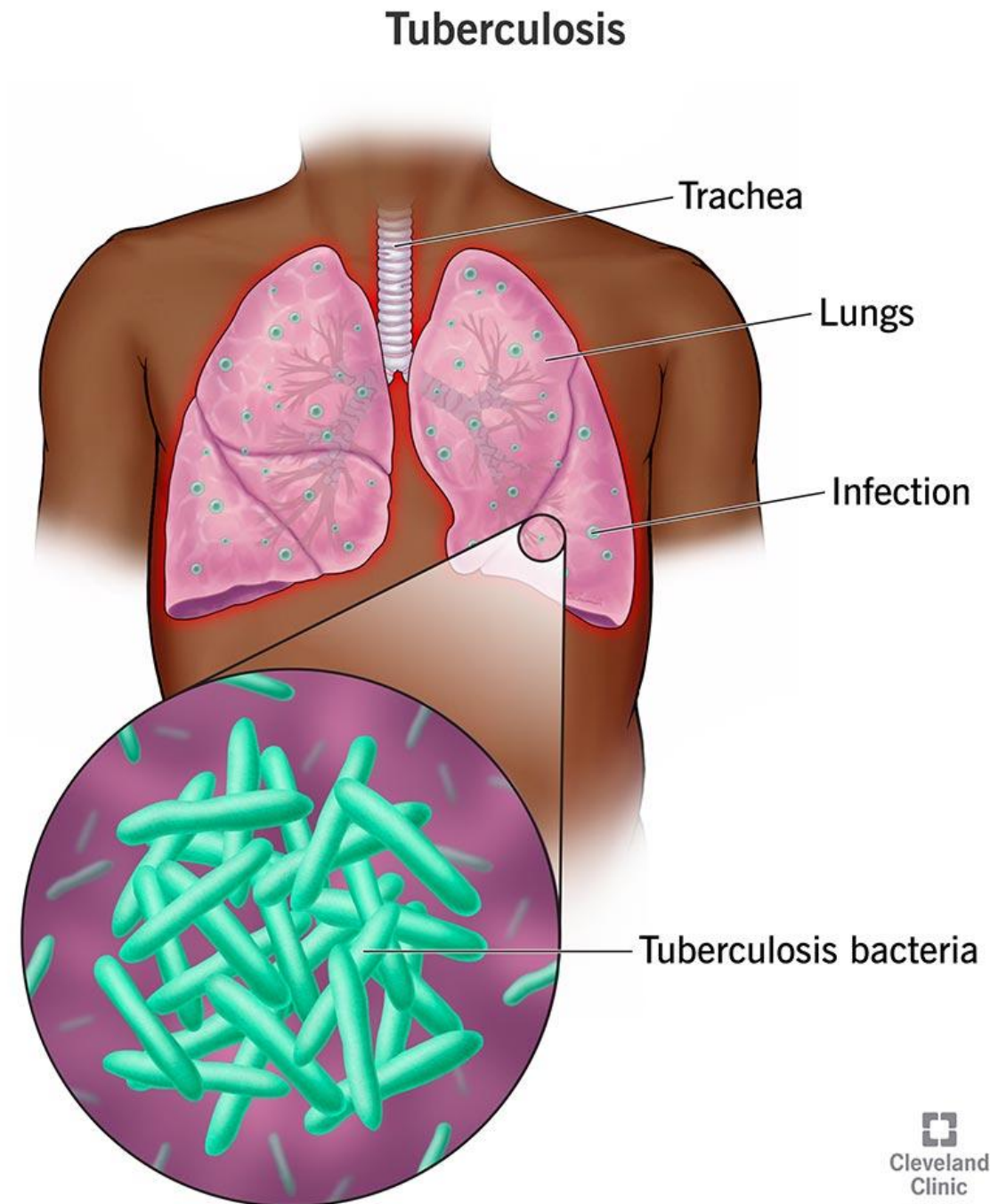
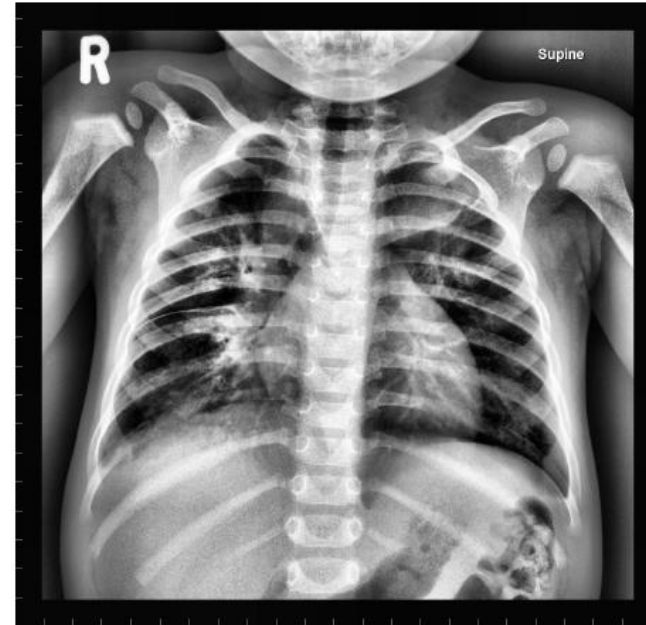
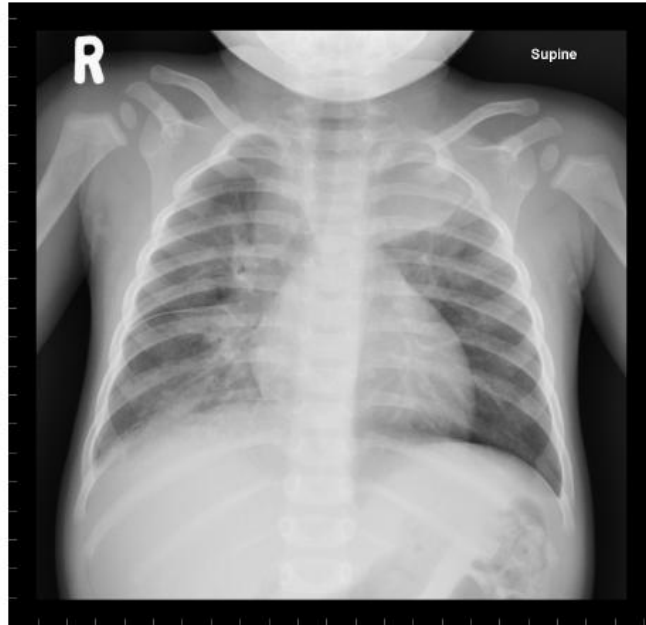
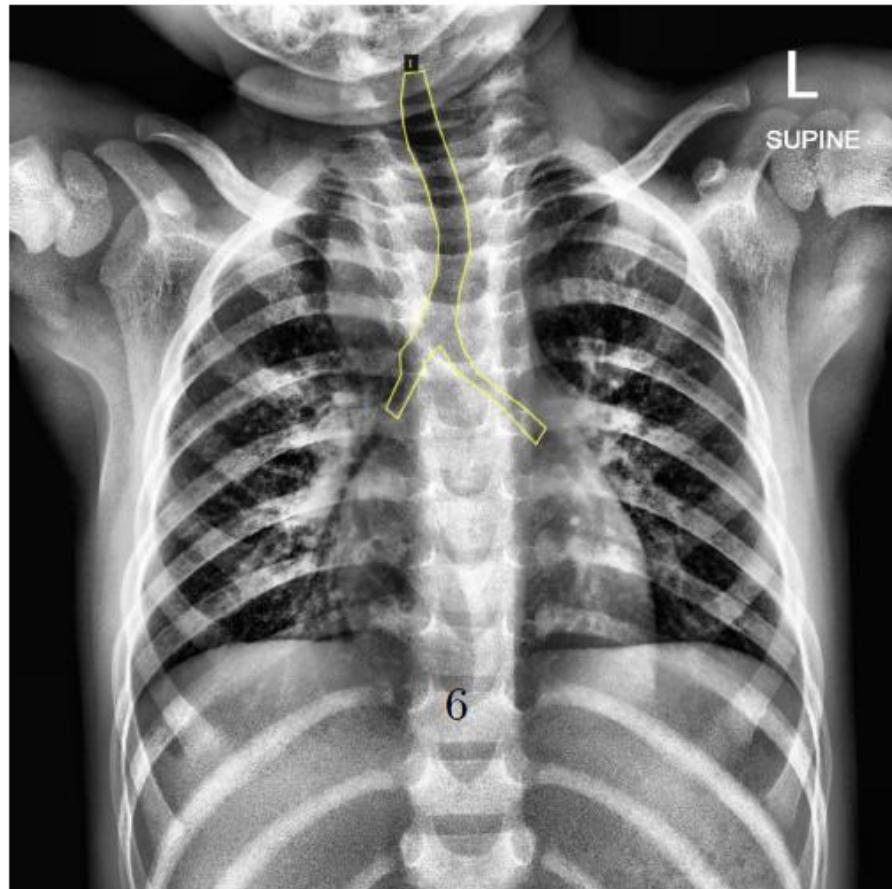


Image enhancement



CXR image examples before and after CLAHE have been applied.

Manual annotation



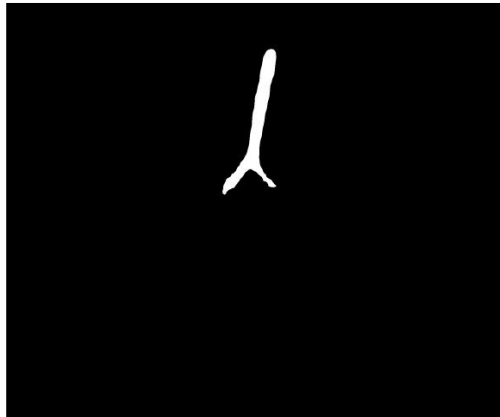
Preprocessing



(a)



(b)



(c)



(d)

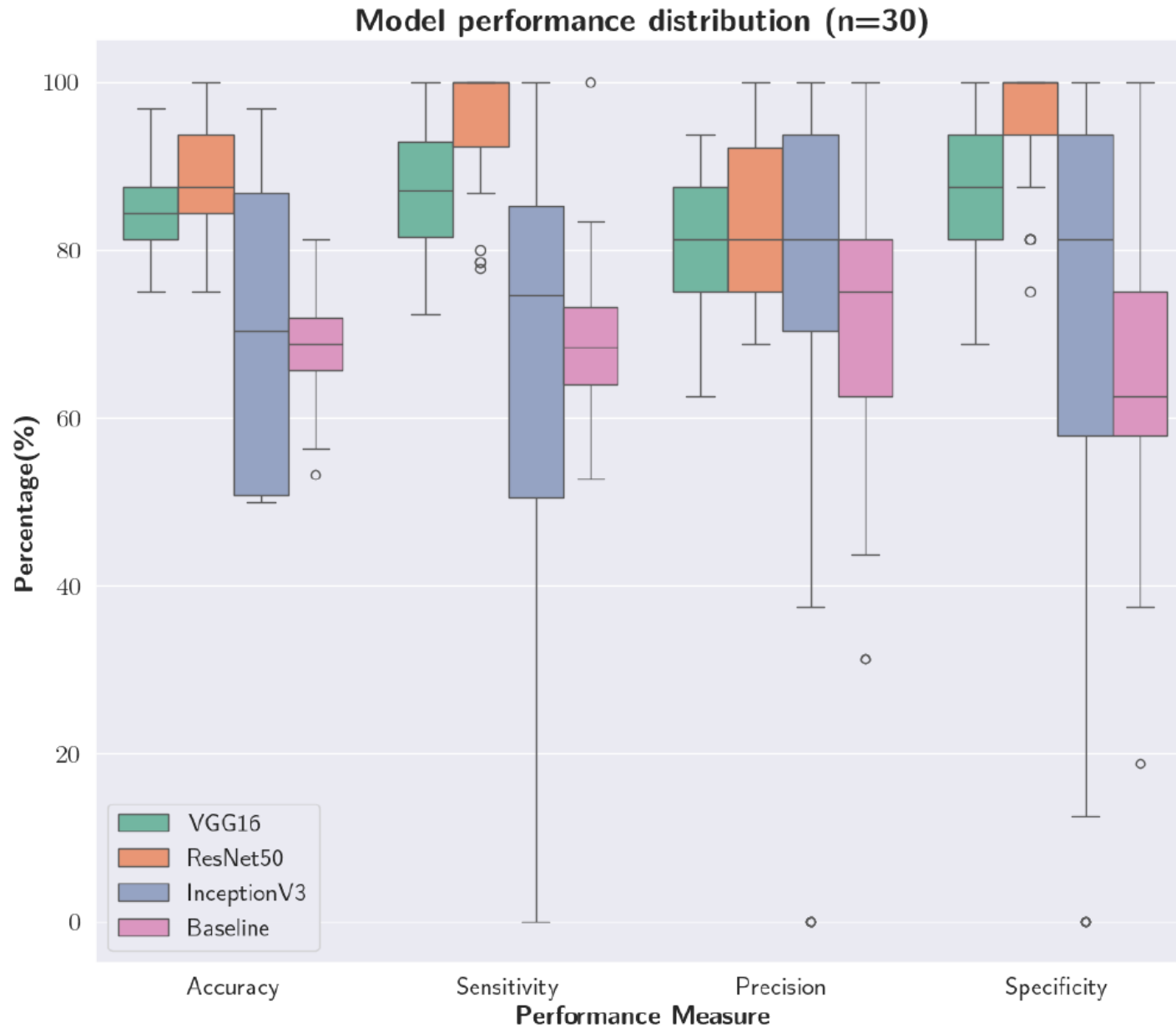
(a) Original CXR Image.

(b) Enhanced image in (a) with the CLAHE algorithm applied.

(c) The resulting mask after the Mask R-CNN model was applied.

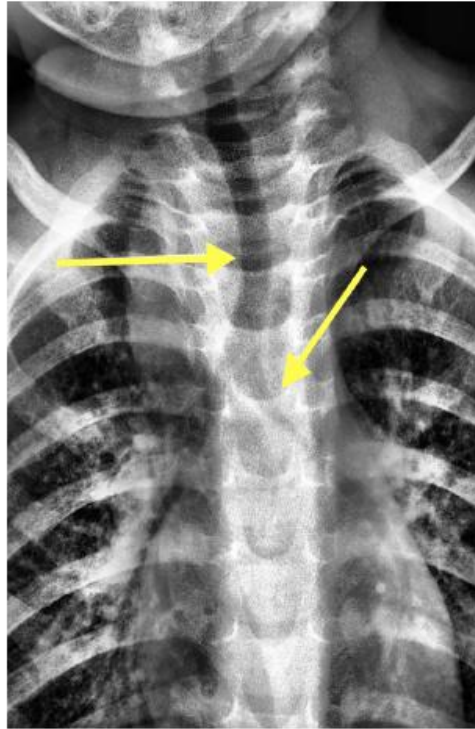
(d) The cropped image of the mask overlayed on the enhanced image.

Results



Recall: 94.43%
Specificity: 94.79%

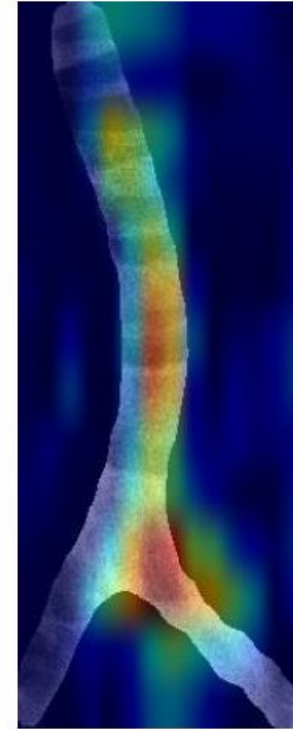
Correctly predicted samples



(a)



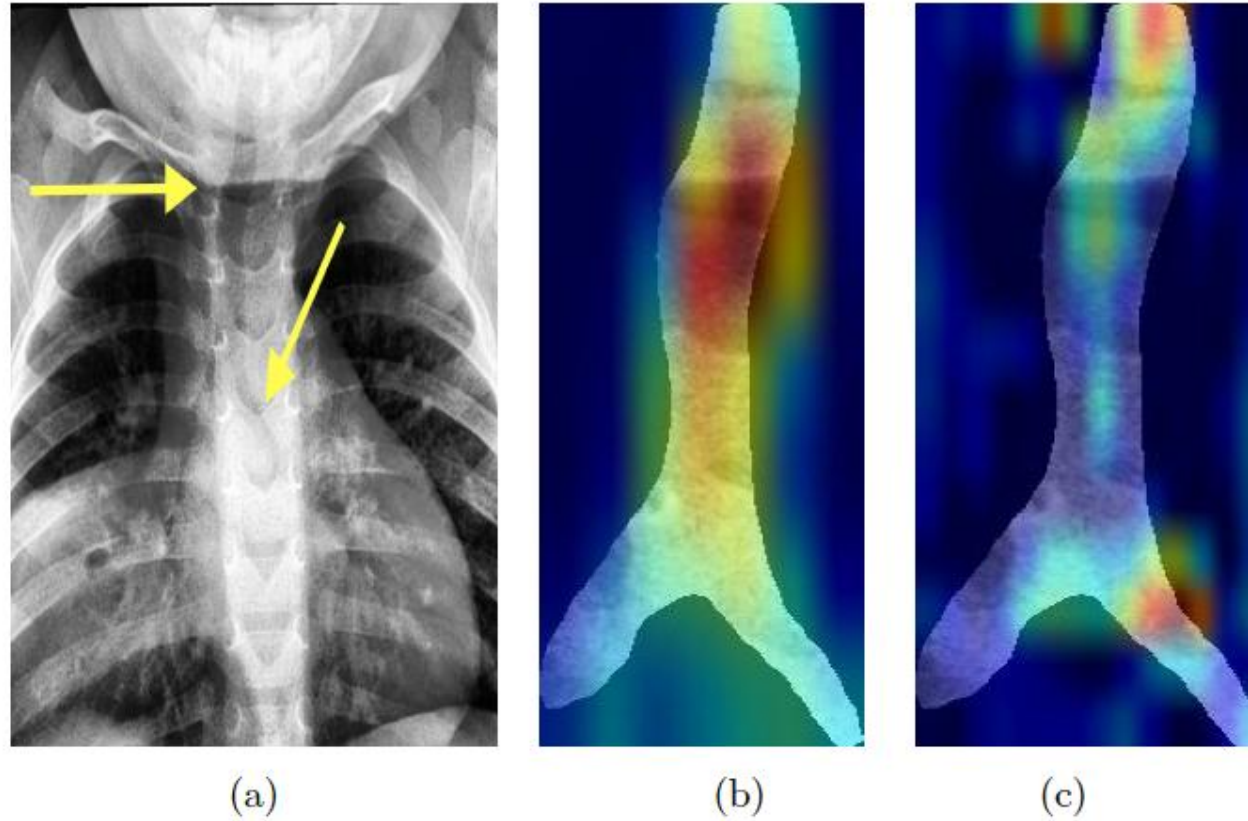
(b)



(c)

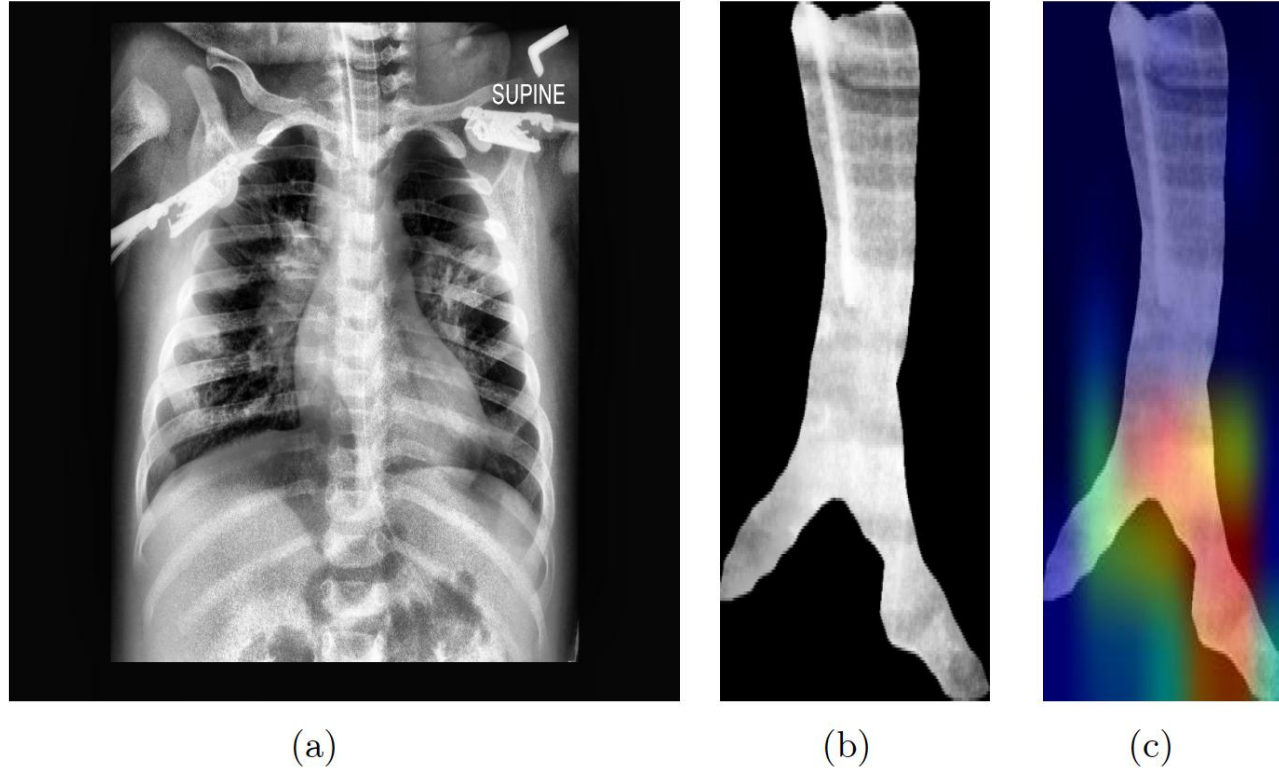
- (a) The enhanced and cropped CXR image where TB is present.
- (b) Grad-CAM heatmap of the custom ResNet50 model.
- (c) Grad-CAM heatmap of the custom VGG16 model.

Correctly predicted samples



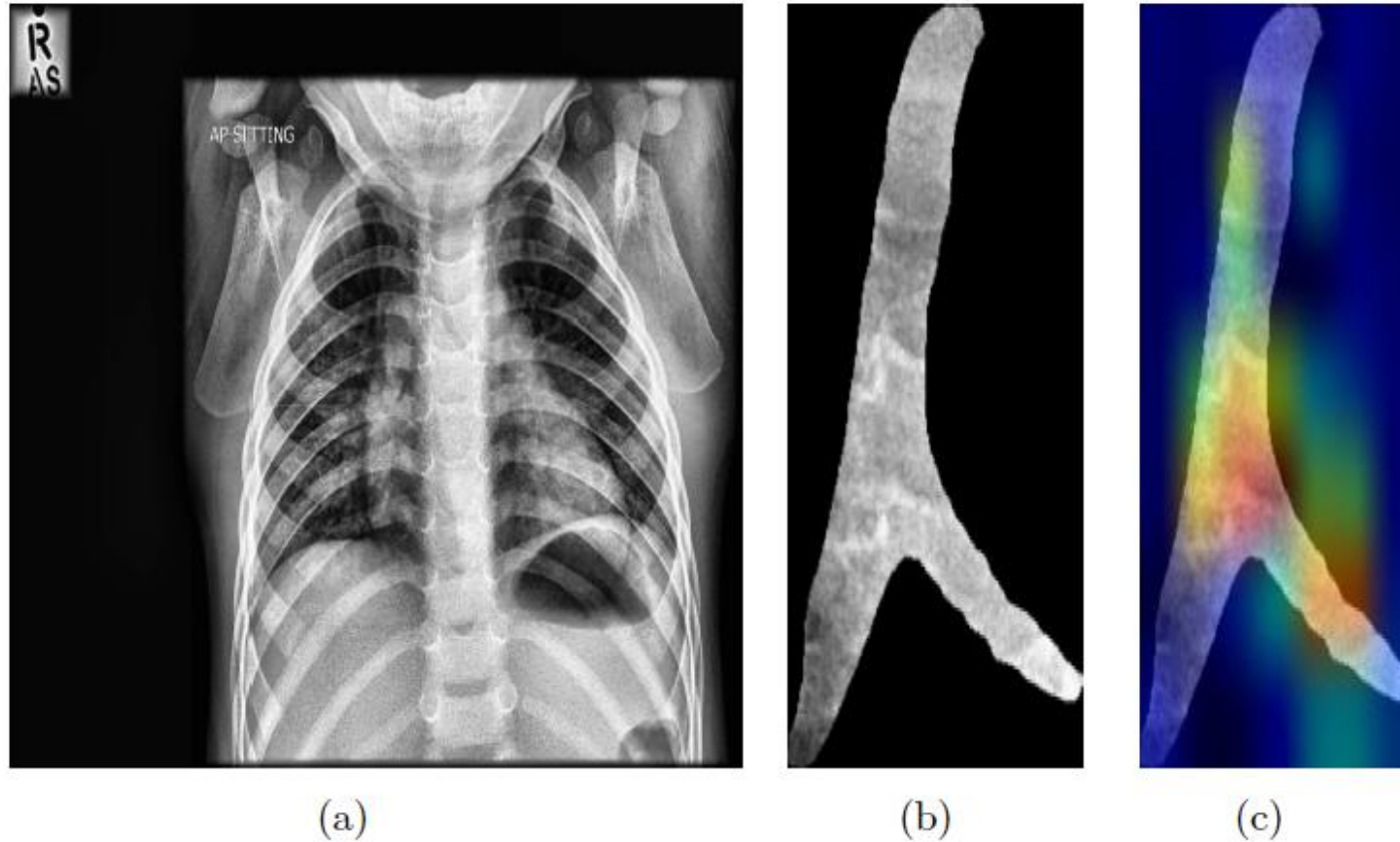
- (a) The enhanced and cropped CXR image where TB is present.
- (b) Grad-CAM heatmap of the custom ResNet50 model.
- (c) Grad-CAM heatmap of the custom VGG16 model.

Normal prediction where TB is present



- (a) Original enhanced CXR image.
- (b) Segmented and enhanced trachea and main bronchi.
- (c) Grad-CAM heatmap of normal prediction activation.

TB predicted where no TB is present



- (a) Original enhanced CXR image.
- (b) Segmented and enhanced trachea and main bronchi.
- (c) Grad-CAM heatmap of TB prediction activation.



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Thank you
Enkosi
Dankie