Deep learning in emergency care

Axel Nyström, 2021-12-09

About me

- MSc in computer science and engineering
- 2 years as data scientist in Malmö

About me

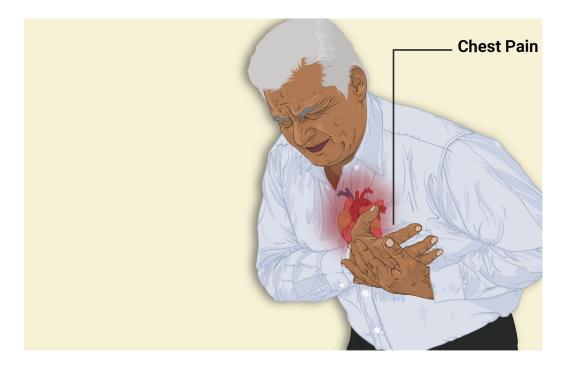
- Machine learning at the emergency department
- Focus on patients with chest pain

About me

- Jonas Björk
- Ulf Ekelund
- Jakob Lundager Forberg
- Mattias Ohlsson

Machine learning for predicting major adverse cardiac events using serial electrocardiograms

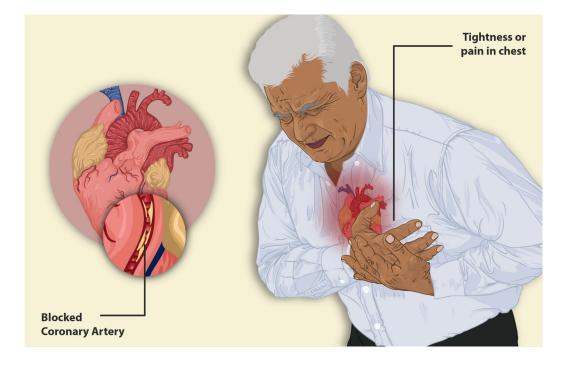
Chest pain



Angina pectoris

Angere = "to strangle"

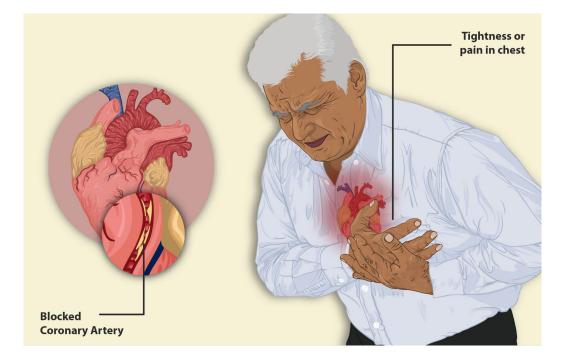
Pectus = "chest"



Angina pectoris

Insufficient oxygen (ischemia)

Can lead to heart attack



What the doctor wants to know

- Send home?
- Send to the cardiac care unit?

What the doctor wants to know

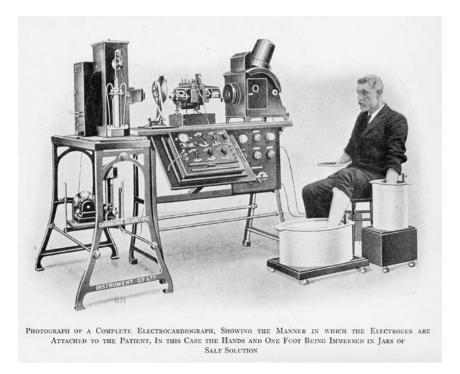
Major Adverse Cardiovascular Event?

- Chest pain
- Heart attack
- Death

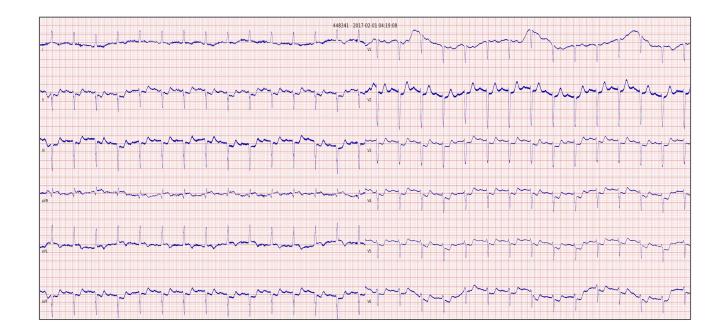
What the doctor does

- Looks at the ECG
- Compares with previous ECGs
- Takes blood sample

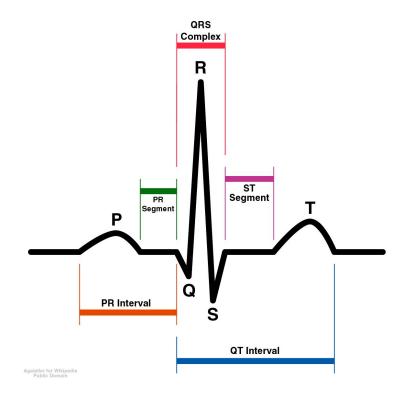
What is an ECG?



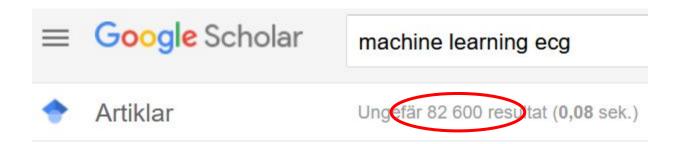
What is an ECG?



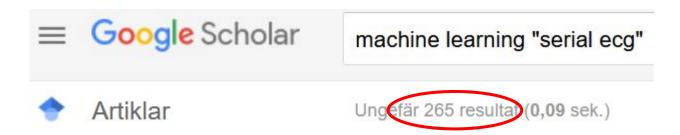
What is an ECG?



Problem formulation



Problem formulation



Problem formulation

Is there any added value of previous ECGs for predicting MACE?

ESC-Trop

- Patients with chest pain at ED
- Consecutive between 2017-2018
- 26547 unique patients



ESC-Trop

- Patients with chest pain at ED
- Consecutive between 2017-2018
- 26547 unique patients
- 26267 non-STEMI
- 26044 with TnT measurements
- 24048 with ECG records at index
- 19500 with >1 ECG record



• Pick a model

- Pick a model
- Tune it for 1 ECG

- Pick a model
- Tune it for 1 ECG
- Tune it for 2 ECGs

- Pick a model
- Tune it for 1 ECG
- Tune it for 2 ECGs
- Compare to see which was better

• What features to include?

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- What model to use?

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What features to include?

Only ECGs

- Simpler
- Easier (maybe)

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Only ECGs

- Simpler
- Easier (maybe)

ECGs + age, sex, troponin

• More clinically relevant

- What features to include?
- What model to use?
- How to tune the model?
- How to quantify the improvement?

Serial ECGs





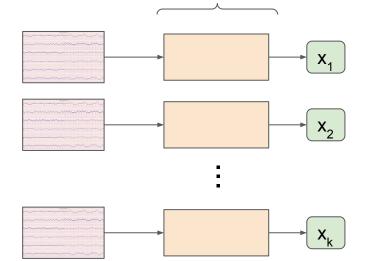
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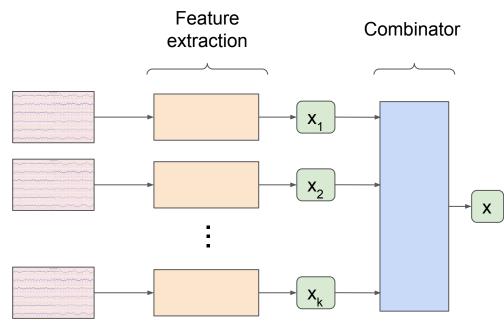
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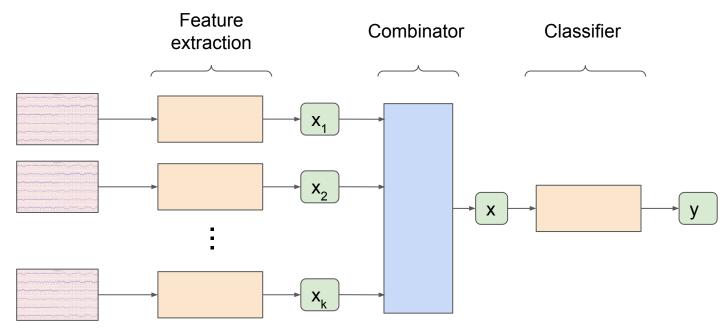
Feature extraction



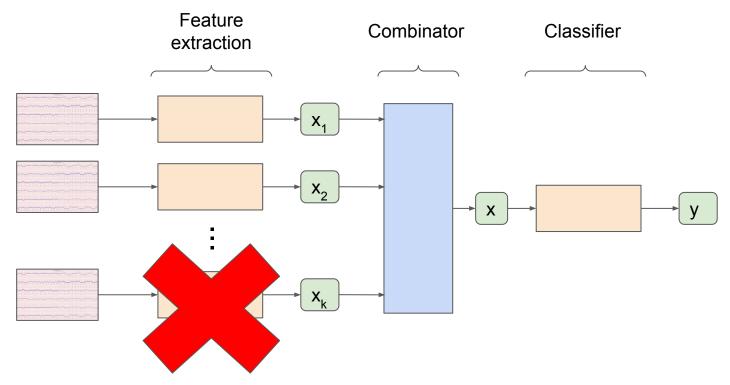




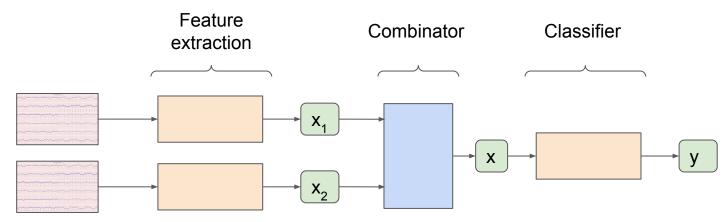
Serial ECGs



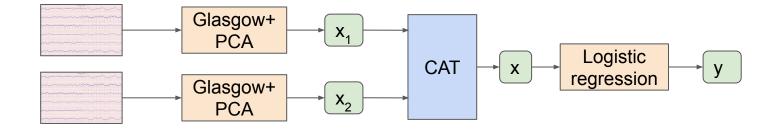
Serial ECGs



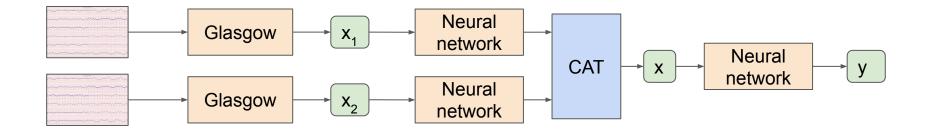
Serial ECGs



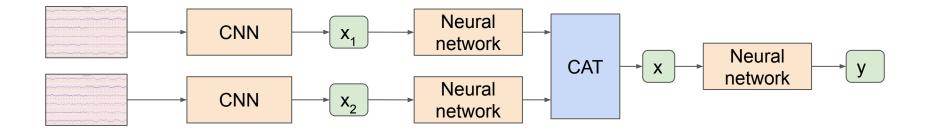
Serial ECGs - Logistic regression



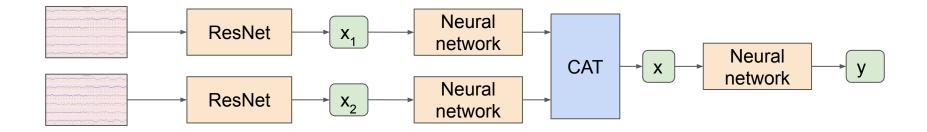
Serial ECGs - Neural Network



Serial ECGs - Convolutional Neural Network



Serial ECGs - Pre-trained ResNet



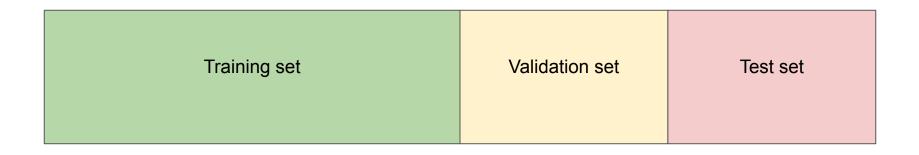
What model to use?

- Logistic regression (LR)
- Neural network (NN)
- Convolutional neural network (CNN)
- Pre-trained ResNet (RN)

Problems

- What features to include?
- What model to use?
- How to tune the model?
- How to quantify the improvement?

How to tune the model?



How to tune the model?

Training set	Validation set	Test set
50% (9750 patients)	25% (4875 patients)	25% (4875 patients)

How to tune the model?

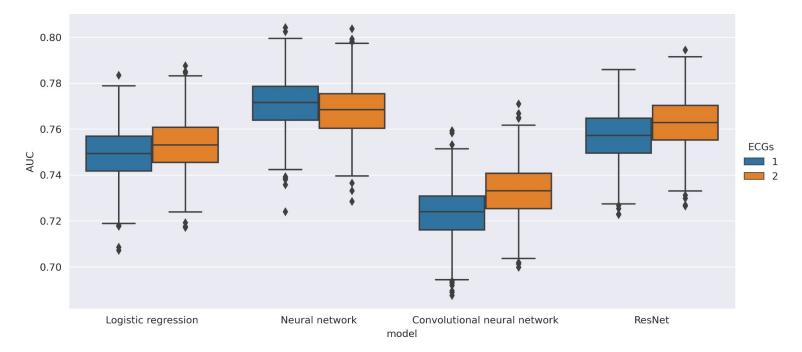
- Train/validation/test set split
- Random search
- Ensemble of best models

Problems

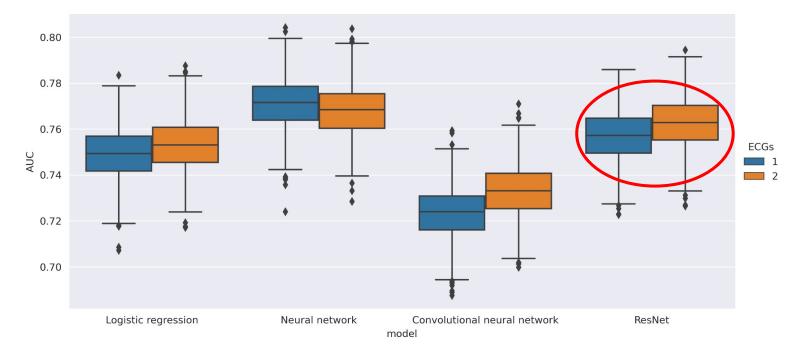
- What features to include?
- What model to use?
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How to quantify the improvement?

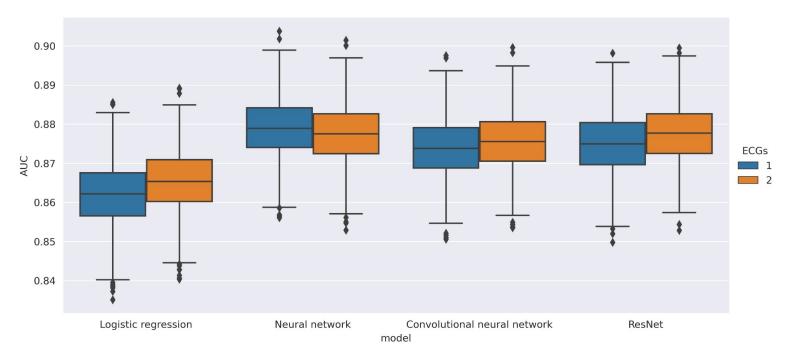
- Metric: ROC AUC
- Bootstrapping



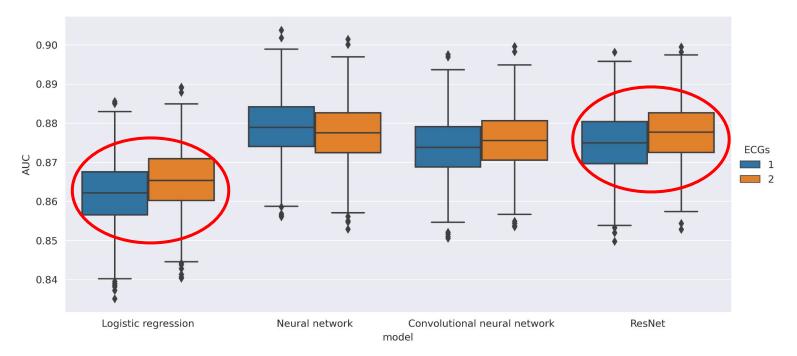
ECGs only, Test data



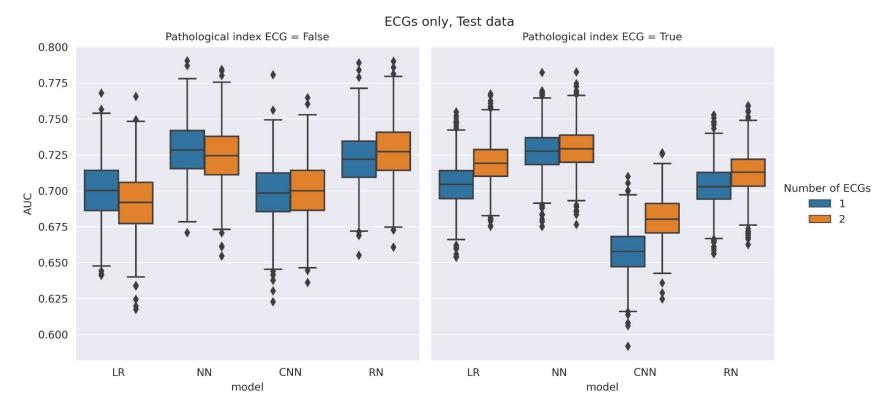
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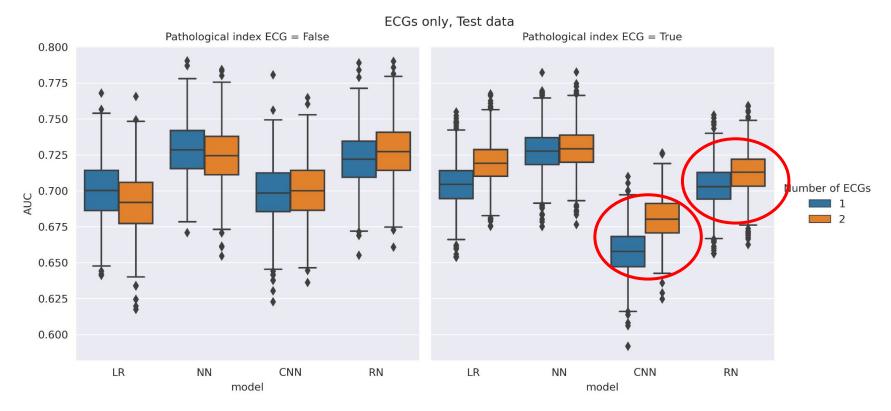


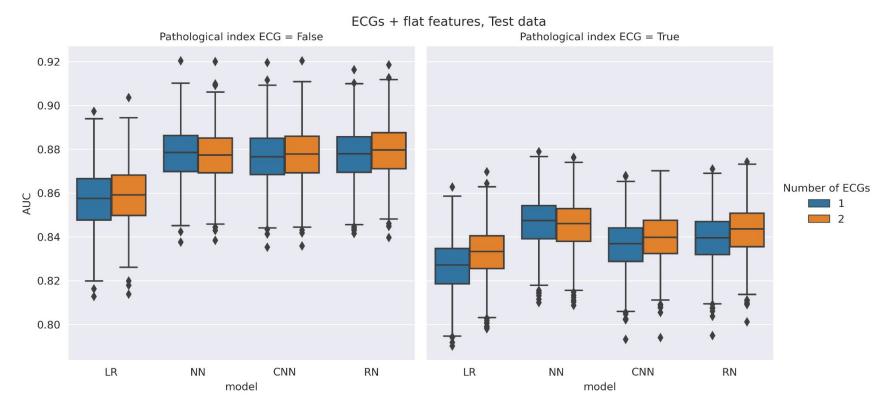
ECGs + flat features, Test data

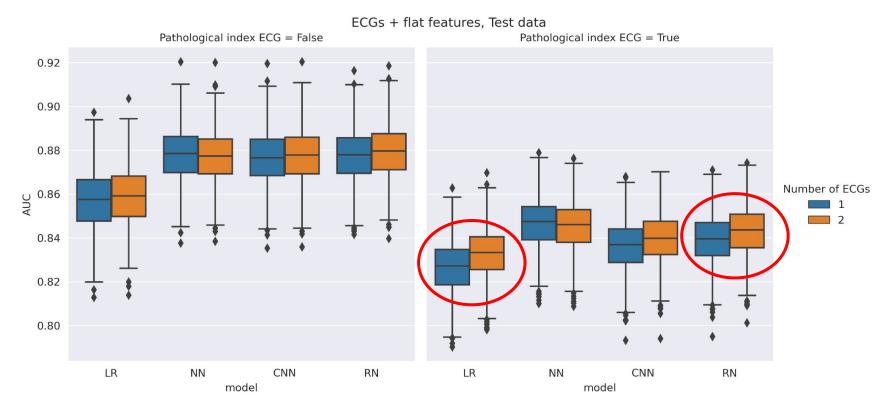


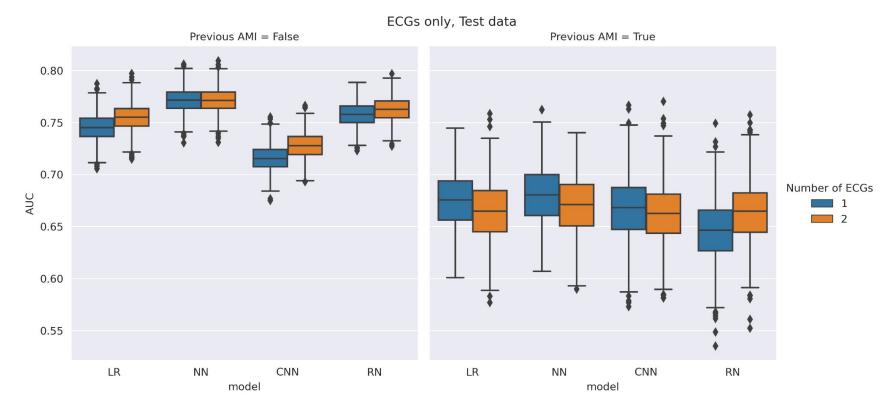
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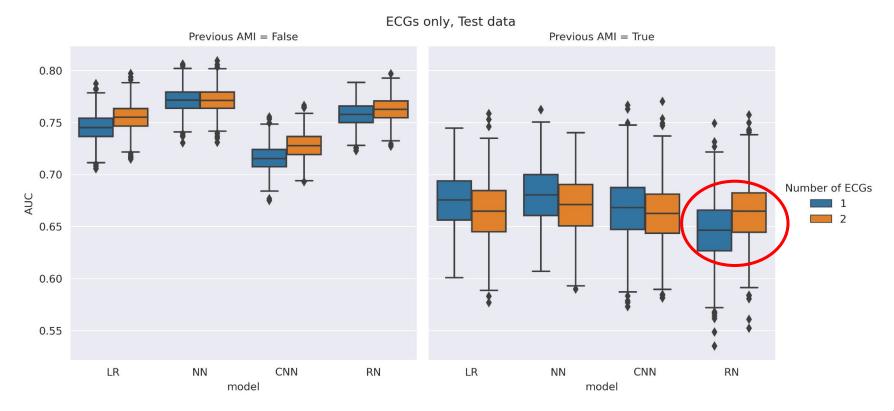


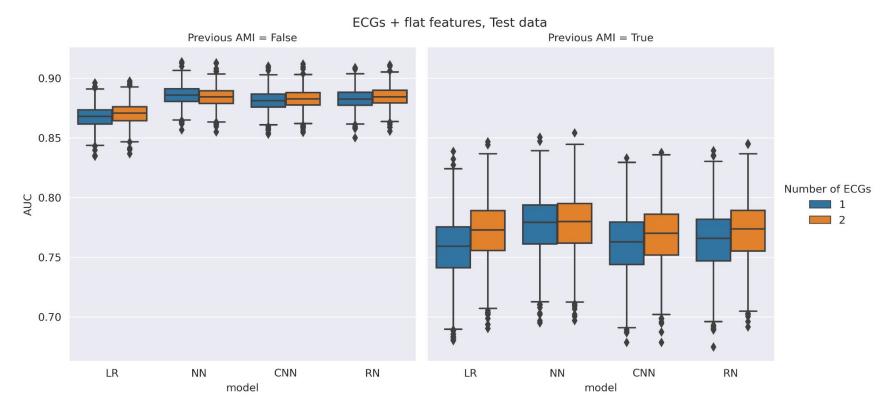


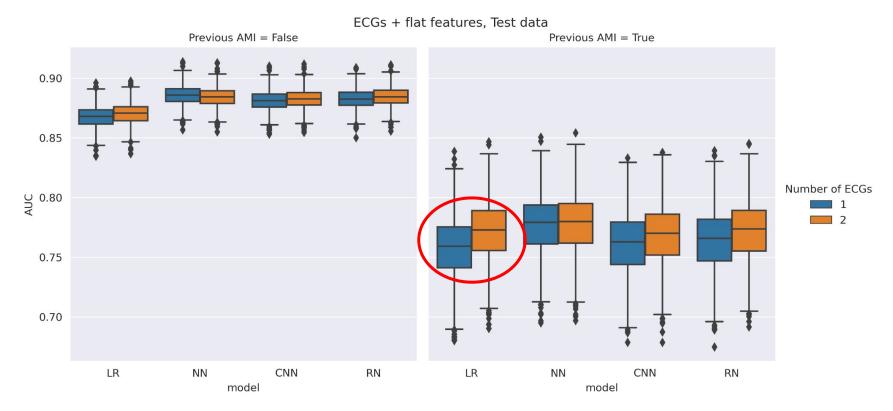












Conclusions

• There might be a signal in the previous ECGs

Conclusions

- There might be a signal in the previous ECGs
- But this signal is weak enough to be practically useless

Questions?