

Vectorcardiography-Based Deep Learning Applied to a Novel Credit Card Sized Device for

Atrial Flutter and Atrial Fibrillation Detection

innovators at heart

Cardiac Electrophysiology
Medical Director of Machine Learning
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Mount Sinai

Fuster

Heart Hospital

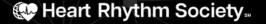
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Disclosures

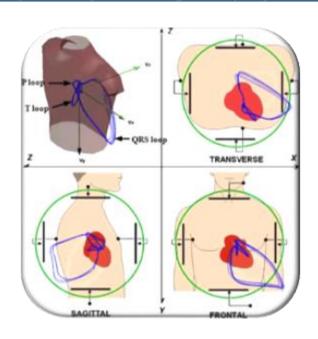
- Consulting: Viz.ai, Heart Sciences (not HeartBeam)
- > Speaking: iRhythm

> Funding: The study was funded by HeartBeam





Background: Vectorcardiography (VCG)



- VCG ~100 years old
- > 3D voltage in x,y,z axes

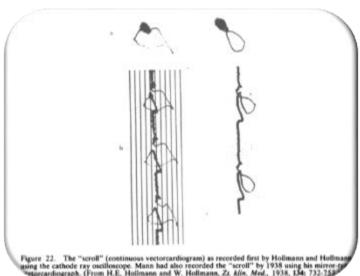
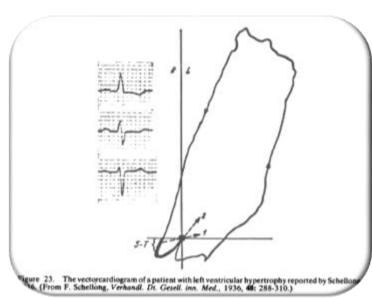
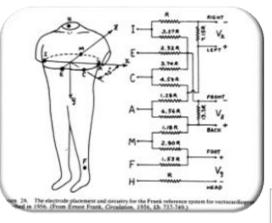


Figure 22. The "scroll" (continuous vectorcardiogram) as recorded first by Hollmann and Hollman using the cathode ray oscilloscope. Mann had also recorded the "scroll" by 1938 using his mirror segregardiograph. (From H.E. Hollmann and W. Hollmann, Zz. kin. Med., 1938, 154: 732-733



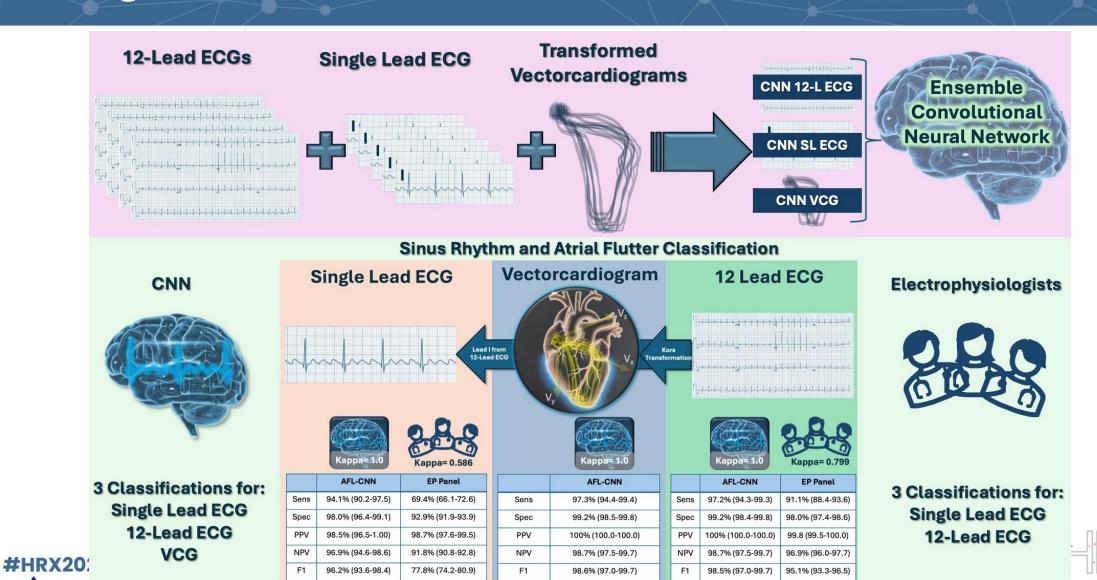




Hasan, Abbott Biomedical Engineering 2015



Background: Transformed VCG CNN





Objective

To assess performance of a convolutional neural network (AFx-CNN) for detecting atrial fibrillation (AF) and atrial flutter (AFL) vs sinus rhythm (SR)

From <u>directly acquired</u> VCGs obtained from a novel credit card sized device (NCCD, Heartbeam)







Methods

201
Consecutive
Patients
Presenting to
EP Clinic



Classification

AFx-CNN Sinus

AF

AFL

- Unblinded 3 electrophysiologist panel as ground truth to define
 - > 131 sinus rhythm (SR) ECGs
 - > 57 AF ECGs
 - ➤ 13 AFL ECGs
 - Expanded 5-EP panel to achieve consensus if initial adjudication was not unanimous (2 ECGs)





Results

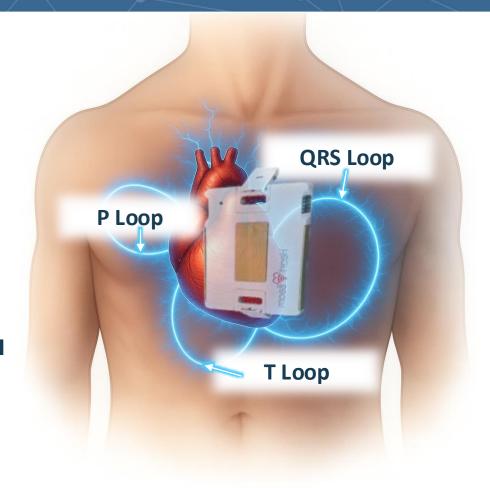
Rhythm	Performance	VCG AFX-CNN (95%CI)	ECG AFX-CNN (95%CI)
	Accuracy	94.5% (91.0-97.5)	95.5% (92.5-98.0)
	Sensitivity	100.0% (100.0-100.0)	98.2% (94.1-100.0)
	Specificity	96.5% (93.2-99.3)	95.1% (91.4-98.5)
Atrial Fibrillation	PPV	91.9% (84.5-98.2)	88.9% (80.6-96.2)
	NPV	100.0% (100.0-100.0)	99.3% (97.7-100.0)
	F1	95.8% (91.6-99.1)	93.3 (88.1-97.5)
	Sensitivity	92.3% (75.0-100.0)	84.6% (61.5-100.0)
	Specificity	97.3% (94.8-99.5)	100.0% (100.0-100.0)
Atrial Flutter	PPV	70.6% (46.7-92.3)	100.0% (100.0-100.0)
	NPV	99.5% (98.3-100.0)	98.9% (97.3-100.0)
	F1	79.5% (60.6-93.8)	91.3% (76.2-100.0)
	Sensitivity	92.4% (87.5-96.7)	95.4% (91.5-98.5)
	Specificity	98.6% (95.2-100.0)	95.4 (91.5-98.5)
Sinus Rhythm	PPV	99.2% (97.4-100.0)	98.4% (95.9-100.0)
	NPV	87.3% (79.5-94.2)	91.9% (85.1-97.4)
	F1	95.6% (92.8-98.0)	96.0% (94.5-98.9)





Conclusions

- The CNN performed similarly, whether applied to 12-lead ECG or <u>directly acquired VCGs</u> in classifying AF, AFL, and SR
- VCG-based cardiac waveform analysis may be advantageous in improving accessibility and scalability
 - > Simplified acquisition
- Further study is warranted to assess whether deep learning algorithms applied to this novel acquisition technology perform comparably to 12-lead ECG for other arrhythmias and disease states









Thank you

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