



# INTELLI 2022



## Applying Deep Learning Techniques in Automated Analysis of Echocardiograms, CMRs and Phonocardiograms for the Detection and Localization of Cardiac Diseases

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**Presented by Racheal Mukisa**





## Short Bio of Racheal Mukisa (Presenter)

- PhD student of Computer Science at Kent State University
- PhD Advisor: Professor Arvind Bansal
- PhD Research area: Applying Deep Learning Techniques to Analyze Cardiac Defects
- Recipient of AAUW (American Association of University Women) Fellowship and Kent State Graduate Assistantship

# Research Interest of the Group

## ■ Social Robotics

- Facial expression analysis
- Gesture analysis
- Gesture generation in humanoids
- Multimodal analysis of human emotions and pain

## ■ Intelligent analysis of Biosignals

- ECG analysis
- Cardiac echogram and MRI analysis using deep learning analysis

## ■ Intelligent analysis of micro-RNA targets to understand human disease such as cancer

# Motivation

- Overall review of the current trends in applying deep learning techniques to understand and localize cardiac defects
  - Analysis of echoCG (echocardiograms) images using CNN, LSTM and their combination
  - Analysis of Cardiac Magnetic Resonance (CMR) images and video frames using CNN, LSTM and their combination
  - Analysis of phonocardiograms (PCG) using CNN and LSTM to detect valvular diseases and fetal cardiac defects
- Automated analysis combining deep learning techniques is a powerful approach due to
  - Non-invasive nature
  - Blood flow related analysis using Doppler echoCG
  - Improved resolution of images with reduced cost, especially for echoCG

# Contribution

- A good comprehensive reference article describing various cardiac diseases involving cardiac muscle defects and valvular defects
- Describes the issues in cardiac chamber and valve detection using image segmentation
- Describes the application of deep learning in identifying
  - Cardiac wall detection
  - Blood-flow volume detection
  - Cardiac wall thickness in ventricle under continuous cardiac motion
  - Wave pattern analysis of PCG



# Background – Anatomy of Heart

## ■ Four chambers:

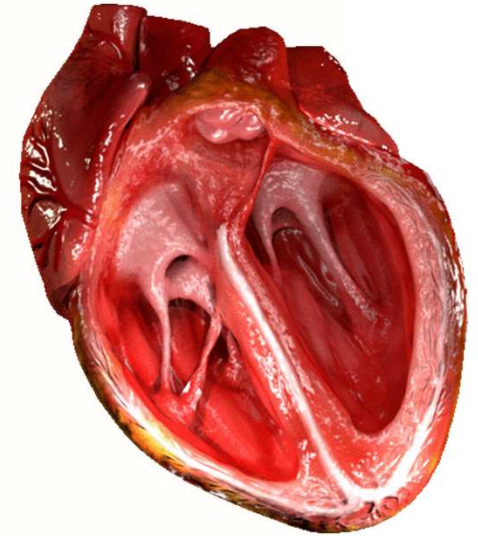
- Right atrium (RA); right ventricle (RV);
- Left atrium (LA); left ventricle (LV)
- Left chambers are separated from right chambers by **septum**
- LV is the biggest chamber

## ■ Blood-flow: **Body** → **RA** → **RV** → **Lung** → **LA** → **LV** → **Body**

- Compression/relaxation occurs due to bioelectric signal flowing in cardiac cells
- Periodic compression of upper and then lower chambers
- Compression in upper chambers moves blood from upper → lower chambers
- Compression in lower chambers pushes blood to lung or body

## ■ Valves to control blood flow

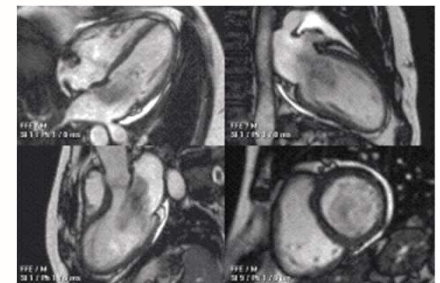
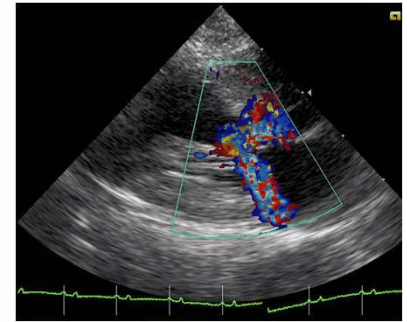
- Tricuspid valve (3 leaflets): RA → RV; pulmonary valve (two leaflets): RV → lungs
- Mitral valve: LA → LV; Aortal valve: LV → body



# Noninvasive Techniques

## ■ Echocardiogram (echoCG)

- Ultrasound technique using reflection and refraction using millimeter waves
- Resolution has improved significantly
- Two types of echograms: TTE (transthoracic echoCG) and transesophageal echoCG (TEE). TTE is completely non-invasive

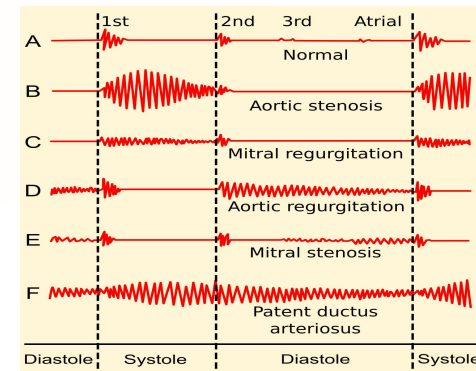


## ■ Cardiac Magnetic Resonance (CMR)

- Uses MRI-based imaging of heart
- Based upon relaxation of magnetically excited hydrogen ions
- Uses multiple 2D slices to simulate 3D images
- Best possible resolution among imaging techniques

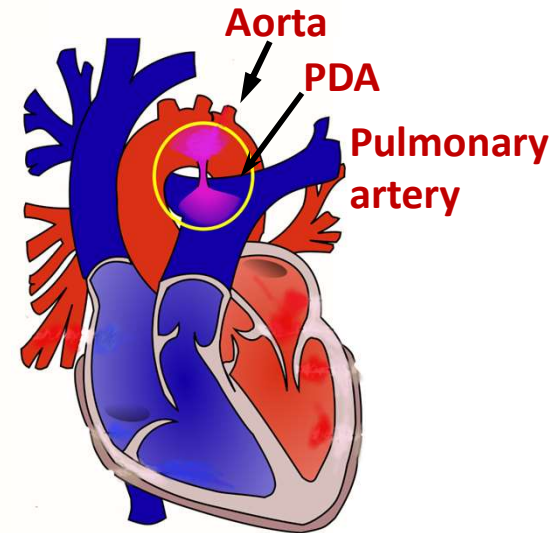
## ■ Phonocardiograms (PCG)

- Sound recordings of continuous opening and closing of valves
- Not suitable for localization of muscle-based and valvular diseases



# Valvular and Heart Muscle Diseases

- Cardiac hypertrophic myopathy
  - Increased muscle thickness and reduced ventricular volume
  - Blood volume flow problem and hypertension
- Valvular diseases causing stenosis and regurgitation
  - Calcification of the leaflets and defects in valves
  - **Ebstein's anomaly**: restricting blood-flow to lungs
- Diseases caused by chamber or blood-channel fusion during birth
  - Blood contamination between de-oxygenated and oxygenated blood
- Perforation or hole in the septum during birth
  - Blood contamination between deoxygenated and oxygenated blood
- Narrowing of blood channels due to deposits causing ischemia, myocardial infarction, or blood-flow problem





# Issues and Advances in Diagnosis

- Invasive investigations for diseases based upon cardiac muscle, chamber or valves are expensive and have side-effects
- ECG can only analyze the presence of arrhythmia, indicate ischemia, myocardial infarction, or electrolyte deficiencies.
  - Cannot localize the diseases caused by defects in cardiac muscle, chamber or valves
- Defects in cardiac muscle, chamber or valves require 3D image analysis, motion analysis and blood flow analysis
- Echocardiogram and CMR provide information about images and motion needed for localization
- PCG provides information about regurgitation and stenosis

# Deep Learning in Diagnosing Cardiac Diseases

Disease Class	Input Mode	DNN Technique
Valvular stenosis + regurgitation	echoCG and PCG	CNN-based segmentation and TGNN + CNN + LSTM
Fetal heart defects	echoCG + CMR	CNN-based segmentation for wall boundaries
Myocardium hypertrophy and myopathy	Doppler echoCG + CMR	Hybrid CNN + LSTM + encoders and decoders for wall thickness, chamber boundaries and blood flow volume
Ischemia and myocardial infarction	CMR	CNN based tissue classification

# Echocardiogram Analysis using Deep Learning

## ■ Used to diagnose

- Stenosis and regurgitation; atrial blockages including atherosclerosis; atrial fibrillation; congenital heart disease; coronary arterial disease; cardio myopathy and hypertrophy; murmur

## ■ Used CNN + LSTM and their variants for segmentation

- Segmentation is measured using non-isotropic diffusion
- Aorta and mitral regurgitation uses region-based CNN, jet-area ratios, jet-stream acceleration at orifice, and PISA
- Area of the orifice is estimated using blood-flow convergence and Doppler's effect

## ■ CHD estimated using multiview echoCG and depth-wise separable convolution

- Filters and combines two separate layers reducing computational complexity
- Multiview provides clarity and evidence from other views

# Issues and Approaches in EchoCG Analysis

## ■ Major problems in chambers identification

- Artifacts
- Discrimination between chambers and septum using similar light intensity
- Fixing missing boundaries between chambers when valves are open

## ■ Multiple variants of CNN-based models

## ■ DW-net derives boundaries precisely using two layers

- Dilated convolution change (DCC) layer collects local and global features
- W-net derives precise boundaries using repeated encoders + decoders

## ■ Echonet derives precise structure and anatomy of heart, blood-flow volume, chamber enlargement, hypertrophy, etc.

- Uses multiple attributes: age, sex, gender, blood flow volume, LV hypertrophy, pacemaker, LA enlargement, etc.

## Deep Learning based CMR Analysis

- Cardiac region is identified using dynamically changing voxel intensities
  - LV is recognized first being the largest chamber
  - Other chambers are recognized later using a heart model
- Dynamic chamber shape estimation
- CMR analysis estimates end-diastolic-volume (EDV), end-systolic volume (ESV), ejection fraction (EF), and myocardial mass
  - Quantification is used to estimate blood-flow and blood-ejection from LV
- Segmentation approaches used for muscle defects and valvular defects
  - Approaches are image driven or model driven
  - Image driven approaches use intensity-based histogram analysis, clustering, region growing, and active contours to identify blood pools
  - Model driven approaches use statistical analysis with atlas or shape contours
  - LV segmentation is used for motion estimation, wall thickness, ischemia and infarction



# Phonocardiogram Analysis

- Analysis is based upon diastolic and systolic phases
- Estimates stenosis, regurgitation, atherosclerotic disease and murmur
- Analysis is done either in time-domain or frequency domain
- Time domain analysis uses wavelet transforms + deep learning
- Frequency domain uses Fourier transforms + deep learning
- Time-growing Neural Network (TGNN) for deep learning
  - Divides the waveform in multiple time segments and diastolic and systolic patterns
  - Combines windowing with neural net-based classification
  - Windowing uses a fixed starting point by growing end-point
  - CNN + LSTM based hybrid model is used for the classification of diseases

# Major Datasets

- **EchoNet Dynamic Database**
  - Includes 10330 four chamber videos with human expert annotations
  - Includes measurements, tracings, ejection volume and cardiac motions
- **Creatis Database contains multimodal 2D and 3D echoCG and CMR images, application software and diagnostics**
  - It contains smaller databases including simulated sequences, manual contouring of LV and RV
- **Harvard Database: CMR 4D blood flow datasets of echoCG and CMR data of 108 patients**
- **Tuft Medical echoCG dataset contains information about disease severity**
- **Heart Database contains 3D CMR segmented LV images validated by clinicians**
- **EMIDEC Database contains CMR LV images of 150 patients with heart attacks**
- **Physionet Database contains nine PCG datasets**
- **Cardiac Atlas Database contains pathological datasets of regional heart shapes**

## Discussion and Conclusion

- Review has provided recent trend in applying deep learning for noninvasive image-based analysis of heart muscles defects and valvular defects.
- The analysis combines information from echoCG images, Doppler effect-based blood-flow analysis, CMR multimedia images for heart chambers and PCG-based sound analysis.
- The integrated deep-learning + image-based analysis can detect
  - Heart chambers detection, motion analysis, heart wall thickness, blood-flow estimates
- Many diseases previously not detected or localized noninvasively can be detected and localized accurately and noninvasively.
- In the coming decade, deep learning-based analysis will become part of effective localization of heart defects.