

Artificial Intelligence in Health Care



Prof. Dr. Eng. Wisnu Jatmiko (SMIEEE, CIQaR, CIQnR, CIMMR)

Professor at Faculty of Computer Science, Universitas Indonesia

Advisory Board IEEE Indonesia Section

Advisory Board Tokopedia-AI Center of Excellent, Fasilkom UI

Advisory Board "Asosiasi Dosen Metodologi Penelitian Indonesia"

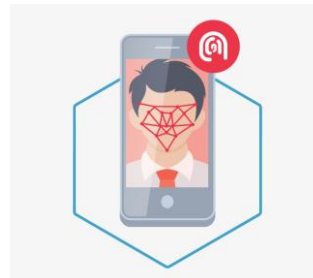
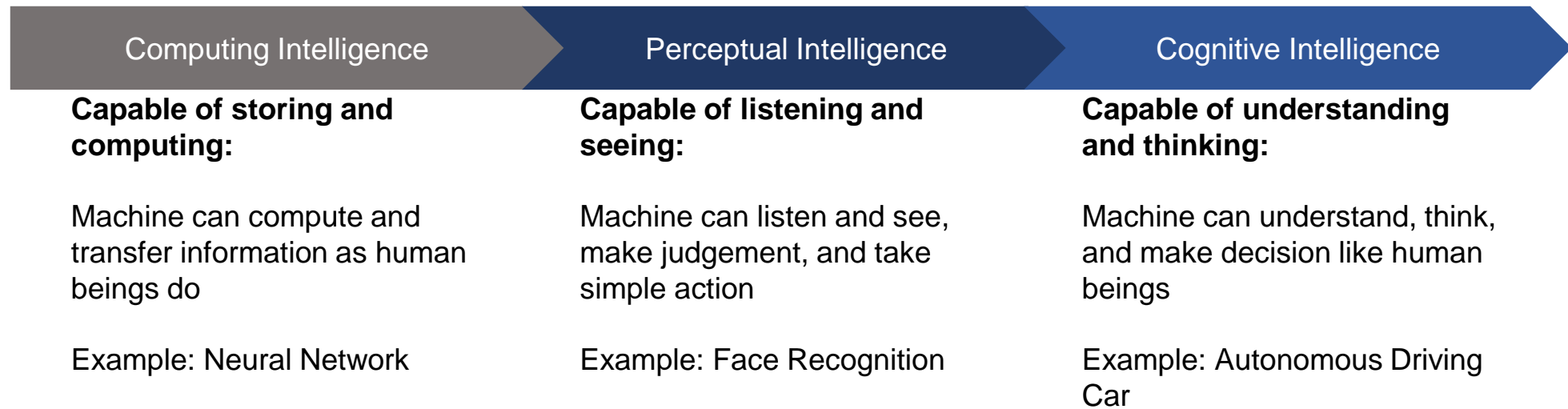
Presentation **Outline**



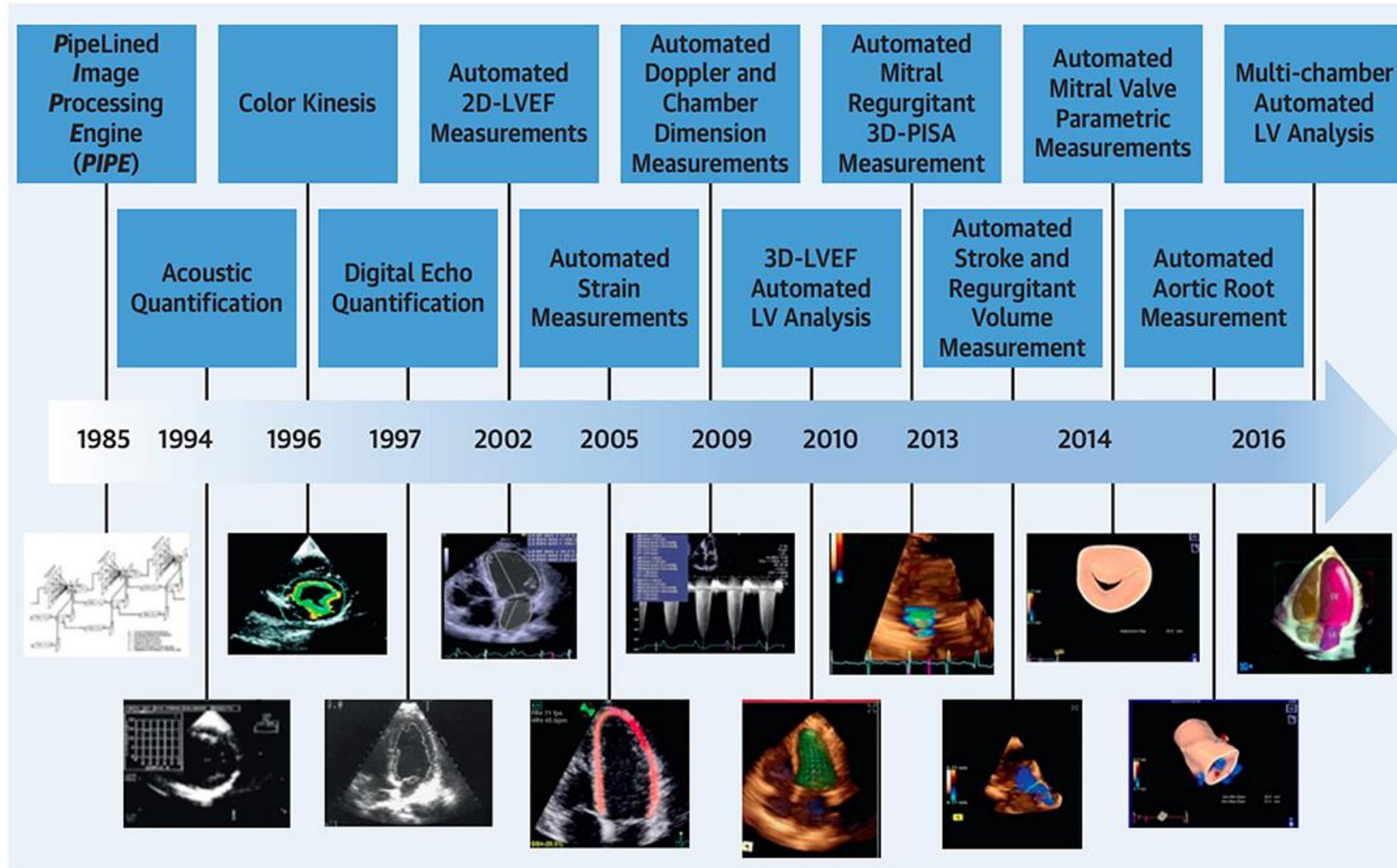
Artificial Intelligence

Definition of AI

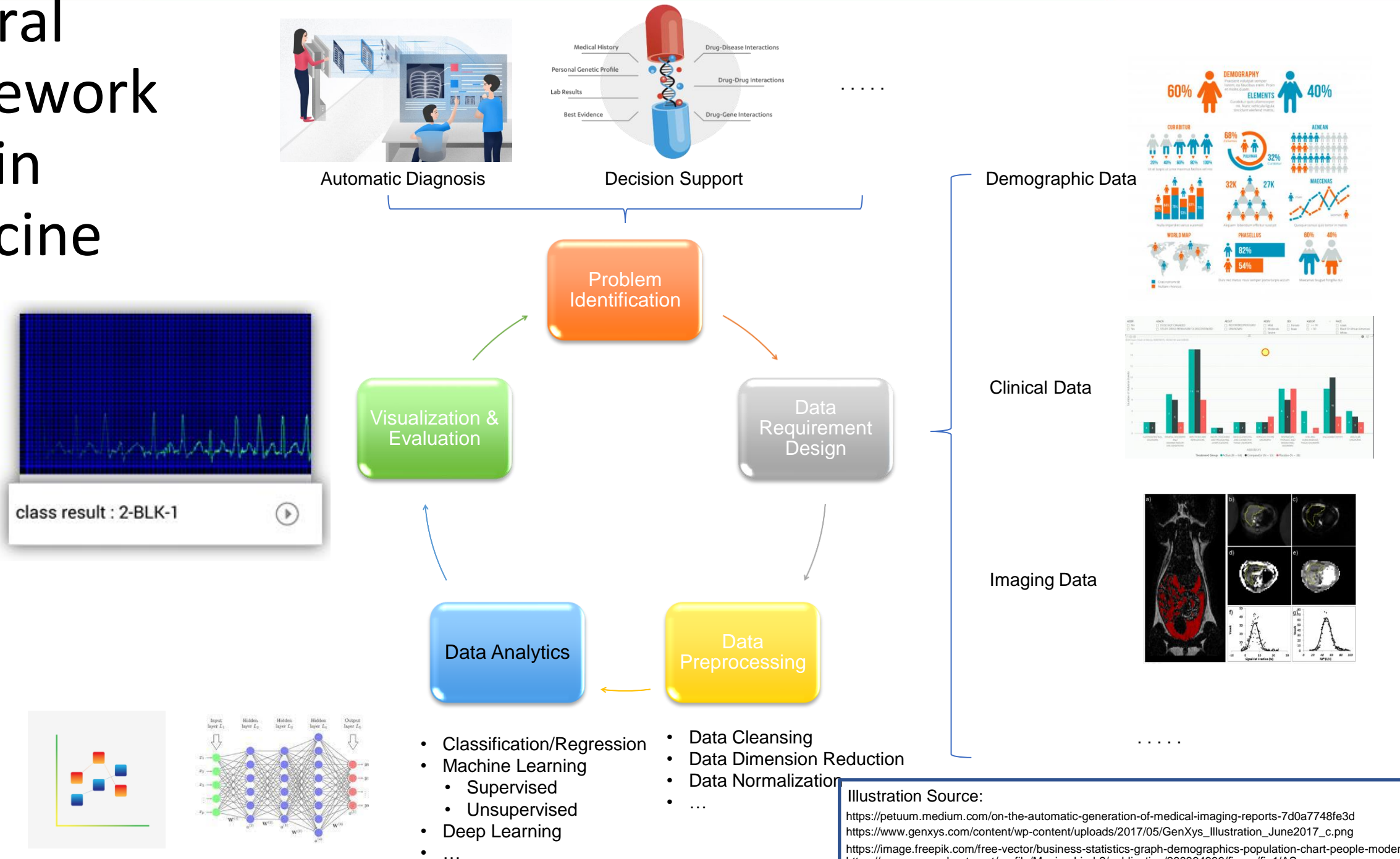
- making a machine behave in ways that would be called intelligent if a human were so behaving. McCarthy, Minsky, Rochester & Shannon, 1956.



Temporal progression in automated quantification in echocardiography. (Adapted from Nolan et al.)

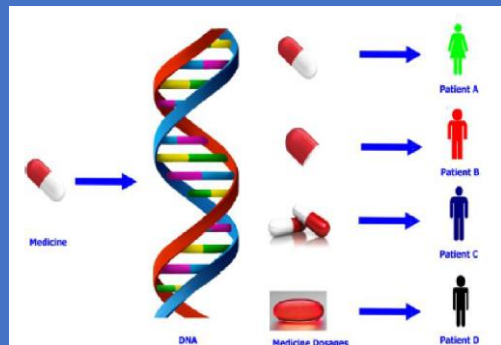


General Framework of AI in Medicine



Problem Identification: Intelligent Healthcare

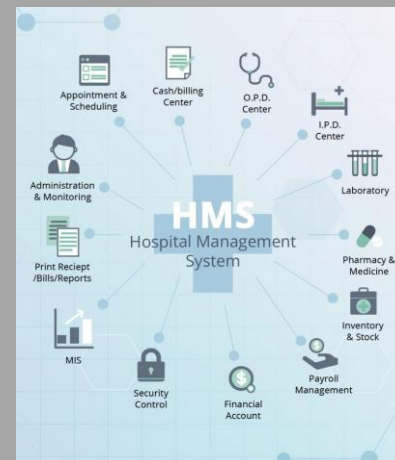
Medicine Mining



Health Management



Hospital Management



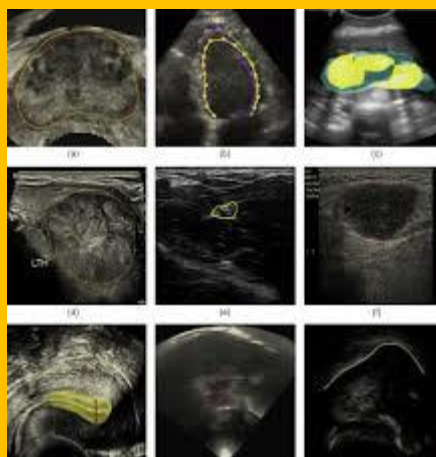
Assistance for Medical Research



Virtual Assistant



Medical Image



Assistance for Diagnosis And Treatment



Disease Risk Forecast

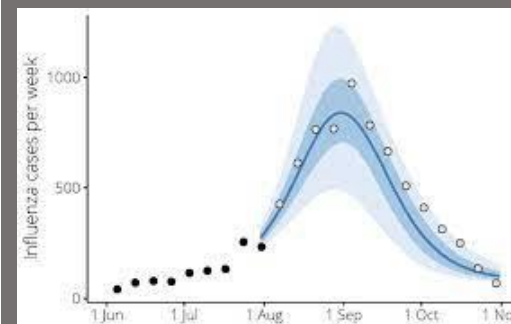


Illustration Source:

<https://techvidvan.com/tutorials/top-8-applications-of-artificial-intelligence-in-healthcare/>
<https://hbr.org/2018/05/10-promising-ai-applications-in-health-care>
<https://petuum.medium.com/on-the-automatic-generation-of-medical-imaging-reports-7d0a7748fe3d>
https://www.12manage.com/images/picture_healthcare_management.png
https://cdn.shortpixel.ai/client/q_lossless,ret_img,w_605,h_708/https://existek.com/wp-content/uploads/....jpg
https://res-4.cloudinary.com/the-university-of-melbourne/image/upload/s--X8x4c9tU--/c_limit,f_auto,q_75,w_890

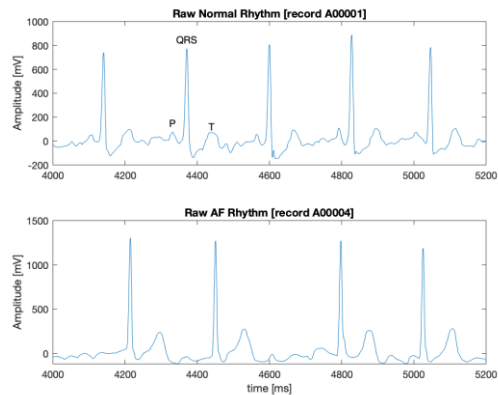
Data Requirement Design: Glossary

- **Dataset** : a collection of data used by our method
- **Sample** : every data record
- **Feature** : Attribute that reflects the condition of a sample
- **Training set**: data used in the training process. The process of forming a model from data is called training / learning
- **Test Set**: The model that has been created needs to be tested using a data set called a test set.

Data Requirement Design: Example of Dataset

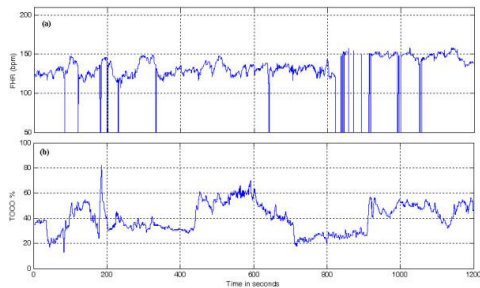
Signals Dataset

ECG Signals



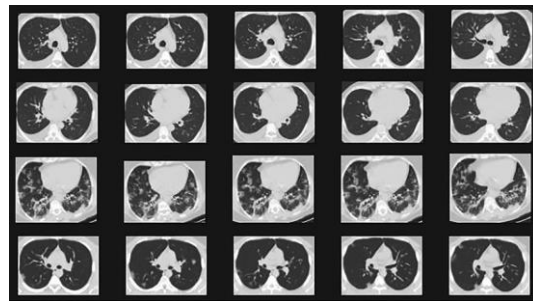
<https://dl.acm.org/doi/fullHtml/10.1145/3421558.3421560>

CTG Signals



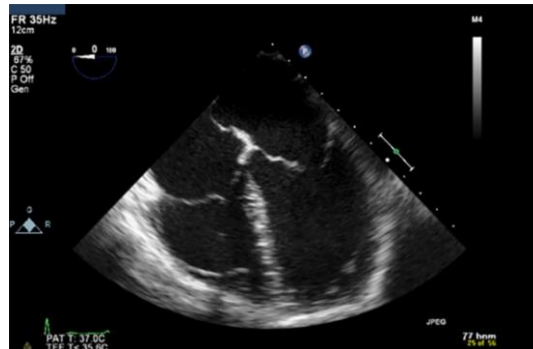
Krupa, Niranjana et.al . (2011). Antepartum fetal heart rate feature extraction and classification using empirical mode decomposition and support vector machine. Biomedical engineering online. 10. 6. 10.1186/1475-925X-10-6.

CT Scan



Chahar, Vijay & Jaiswal, Aayush & Gianchandani, Neha & Singh, Dilbag & Kaur, Manjit. (2020). Classification of the COVID-19 infected patients using DenseNet201 based deep transfer learning. Journal of biomolecular Structure & Dynamics. 10.1080/07391102.2020.1788642.

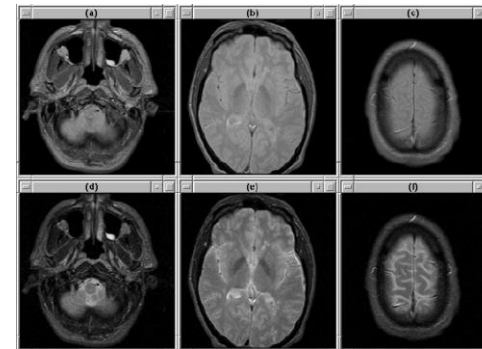
Echo



<https://erp.bioscientifica.com/view/journals/echo/2/4/G29.xml>

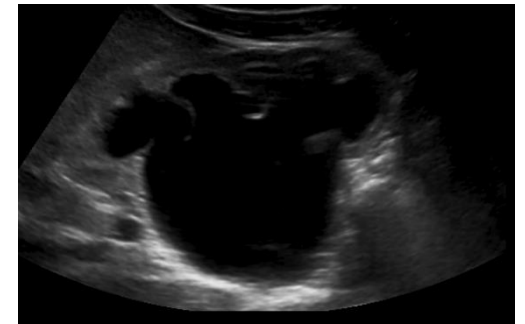
Image Dataset

MRI



<https://www.cs.sfu.ca/~stella/papers/blairthesis/main/node48.html>

USG



Sivanesan, Umaseh & Braga, Luis & Sonnadara, Ranil & Dhindsa, Kiret. (2019). Unsupervised Medical Image Segmentation with Adversarial Networks: From Edge Diagrams to Segmentation Maps.

Type of Data:

Tabular: Medical records, Demographic, Clinical
Signals: ECG Signals, CTG Signals, EEG Signals, etc.
Images: CT Scan, USG, 3D CT Scan, etc.
Videos: Echocardiography video, USG video, etc.

Data Preprocessing:

Good AI is built on Good Data

Data Cleansing

Fill in missing values,
and detect and
eliminate cause of
dataset exceptions

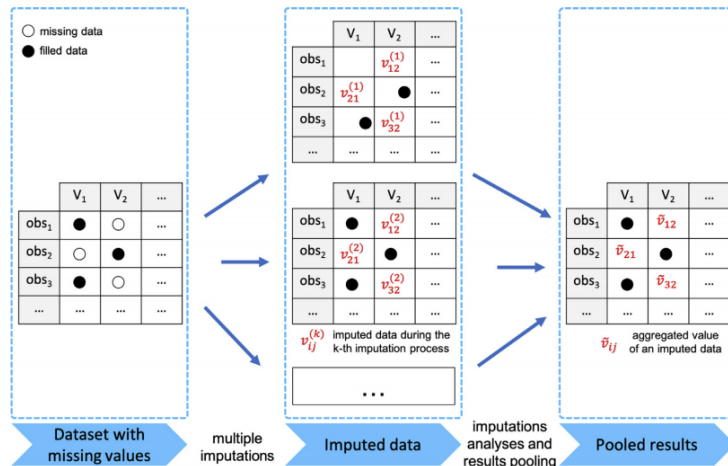
Data Normalization

Normalize data to
reduce noise and
improve model
accuracy

Data Dimension Reduction

Simplify data
attributes/features
to avoid dimension
explosion

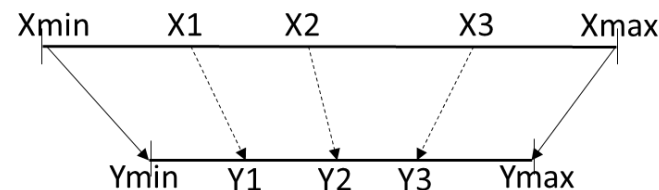
Missing value imputation illustration



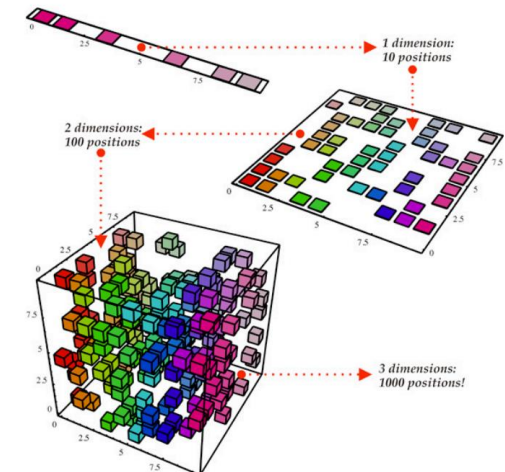
Data normalization illustration

Min, Max Normalization

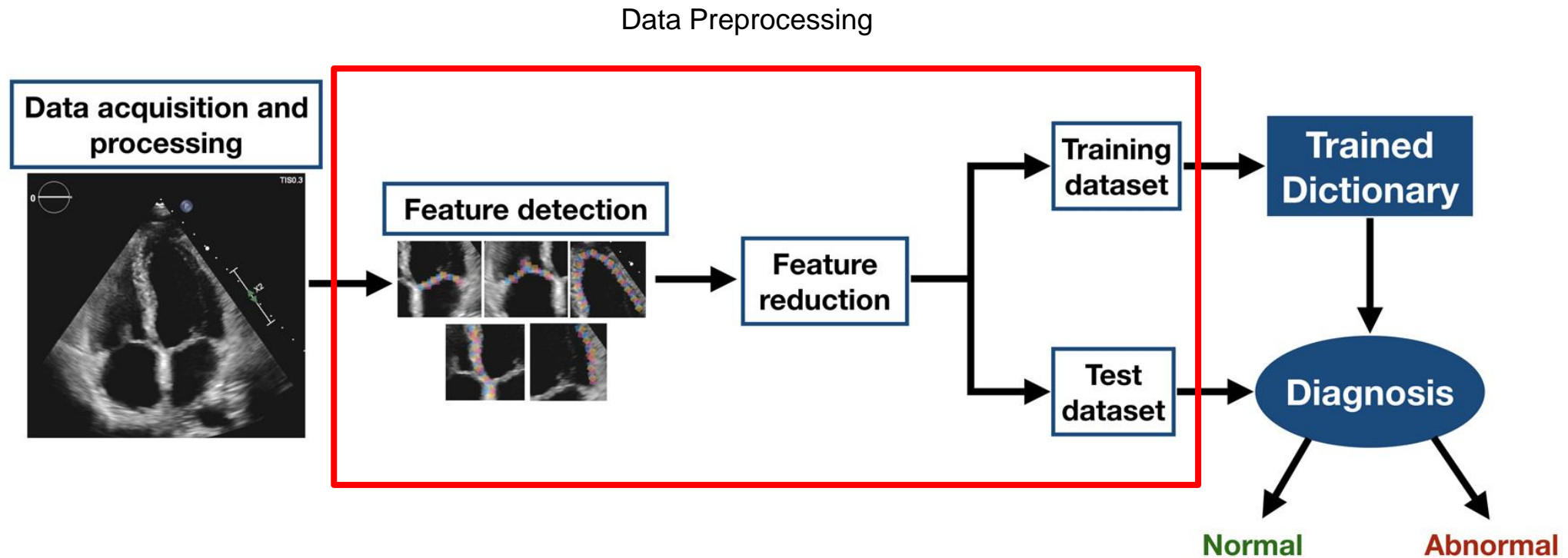
$$X_{new} = \frac{X - X_{min}}{X_{max} - X_{min}}$$



Dimension reduction illustration



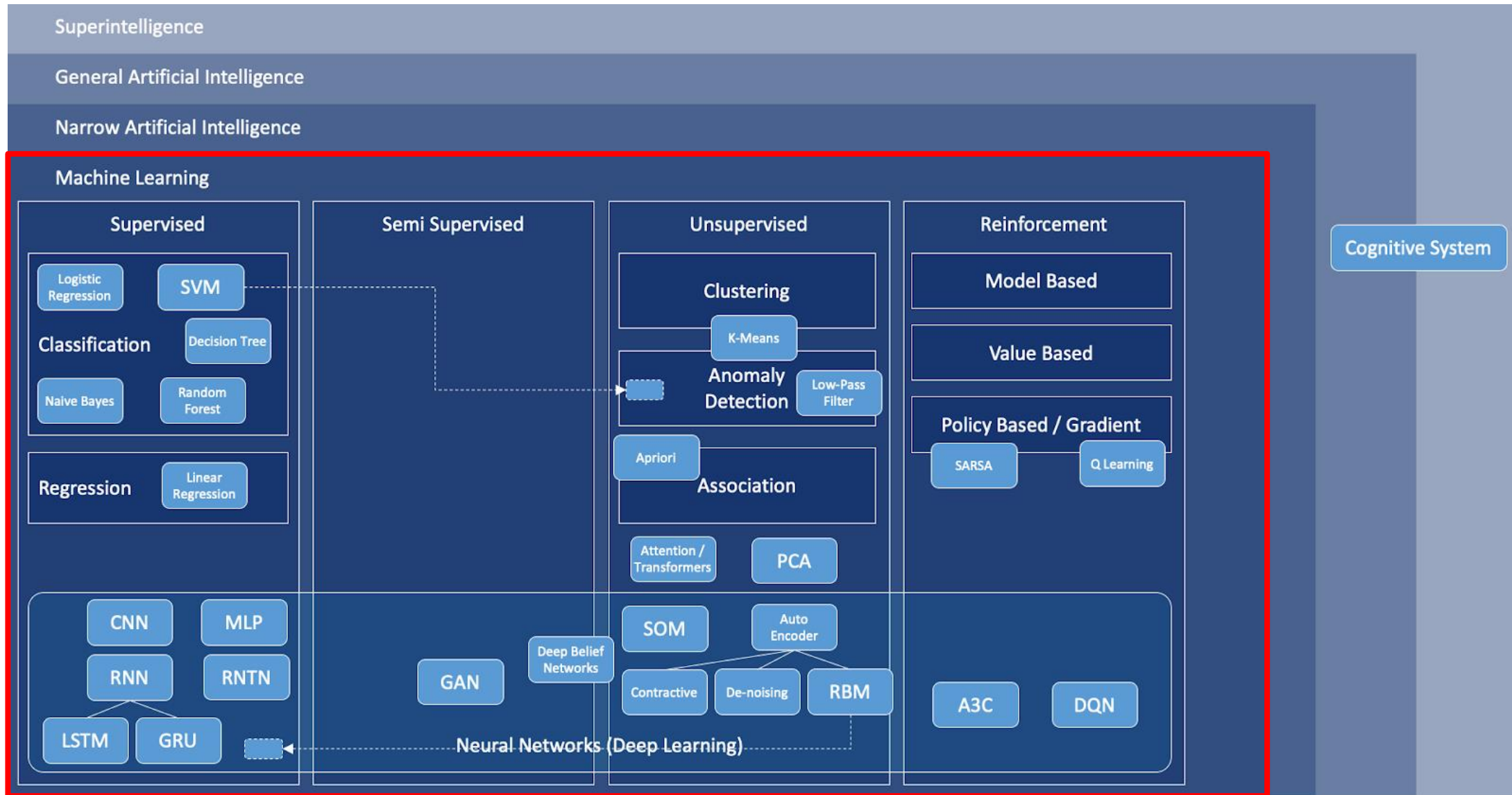
Example for Data Preprocessing



Alsharqi, M., Woodward, W. J., Mumith, J. A., Markham, D. C., Upton, R., & Leeson, P. (2018). Artificial intelligence and echocardiography, *Echo Research and Practice*, 5(4), R115-R125. Retrieved Jun 4, 2021, from <https://erp.bioscientifica.com/view/journals/echo/5/4/ERP-18-0056.xml>

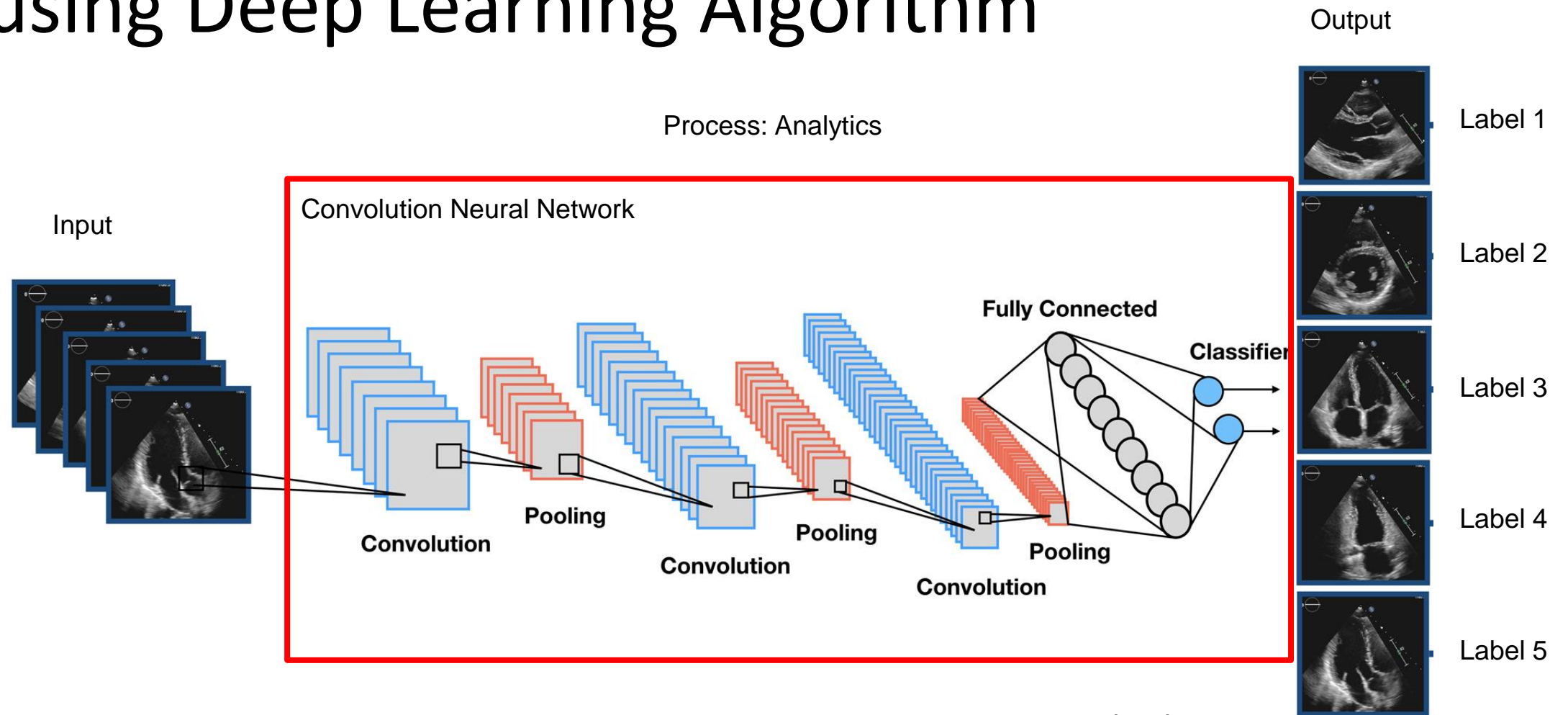
Data Analytics:

"AI and machine learning high level overview" by Nils Ackermann is licensed under Creative Commons [CC BY-ND 4.0](https://creativecommons.org/licenses/by-nd/4.0/)



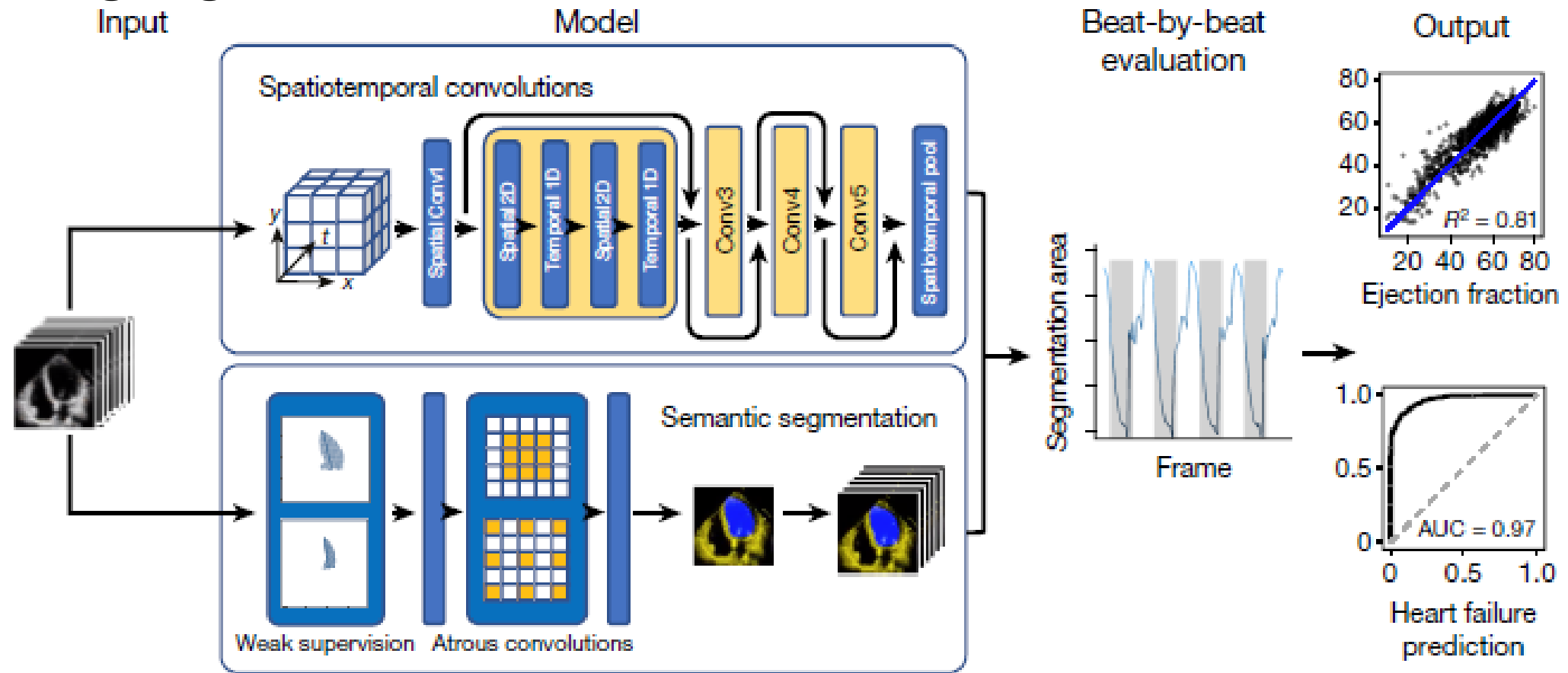
Abbreviations: SVM – Support Vector Machine, CNN – Convolutional Neural Network, RNN – Recurrent Neural Network, LSTM – Long Short-Term Memory, GRU – Gated Recurrent Units, MLP – Multilayer Perceptron, RNTN – Recursive Neural Tensor Network, GAN – General Adversarial Network, PCA – Principal Component Analysis, SOM – Self-Organizing Map, RBM – Restricted Boltzmann Machine, SARSA – State-Action-Reward-State-Action, DQN – Deep Q Network, A3C – Asynchronous Advantage Actor Critic

Example for Echocardiography Classification using Deep Learning Algorithm



Alsharqi, M., Woodward, W. J., Mumith, J. A., Markham, D. C., Upton, R., & Leeson, P. (2018). Artificial intelligence and echocardiography, *Echo Research and Practice*, 5(4), R115-R125. Retrieved Jun 4, 2021, from <https://erp.bioscientifica.com/view/journals/echo/5/4/ERP-18-0056.xml>

Example for Echocardiography Segmentation using Deep Learning Algorithm : EchoNet-Dynamic Algorithm



The model first predicts the ejection fraction for each cardiac cycle using spatiotemporal convolutions with residual connections generates frame level semantic segmentations of the left ventricle using weak supervision from expert human tracings. These outputs are combined to create beat-to-beat predictions of the ejection fraction and to predict the presence of heart failure with reduced ejection fraction. AUC, area under the curve.

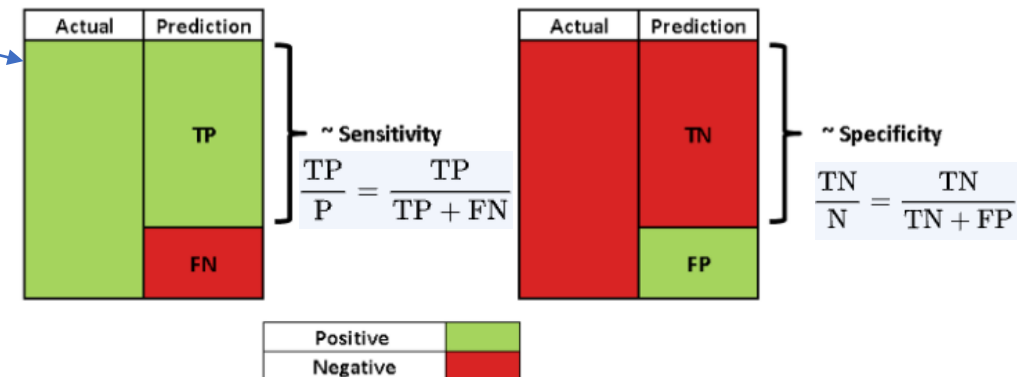
Visualization and Evaluation

Classification/Prediction Problem:

- Classification accuracy
- Confusion Matrix
- Precision and Recall
- F1 Score
- Sensitivity and Specificity
- ROC and AUC
- Etc.

$$Accuracy = \frac{\text{Number of correct predictions}}{\text{Number of all predictions}}$$

Confusion matrix for binary classification			
Actual value	A	TP	FN
	B	FP	TN
		A	B
		Predicted value	



Barriers to AI implementation



Large-scale, labeled datasets with high quality CT image data are required for training and testing new algorithms



Bias can arise in AI algorithms over time, through learning from disparities in patient demographics or healthcare systems



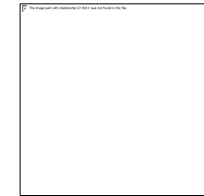
AI algorithms can **often be viewed as “black boxes”** which autonomously learn and make decisions



There is **the legal consideration of clinical clearance** for AI powered software applications.

Future Implications

AI will be integrated into “the system” and **improve the efficiency of clinical workflow.**



AI can be used **to predict** diseases that will occur in the future.



AI can **be able to guide patient therapy** considering clinical factors and other supporting data to form **an accurate decision support**





OUR PROGRESS OF AI IN MEDICINE

Patient in house or everywhere
Takes vital signs and ECG surveys
tailored to the **Self Monitoring
Device**

Infrastructures
With a special algorithm, the server process
clinical data into information. Managed with
two directional data communication with
electronic health records.



Heart beat Classification

ECardio

E-Cardio is an integrated system that helps people to examine their cardiovascular health, without having to meet a doctor. This is especially useful in a situation like Indonesia.

1

Sensors

The system utilizes sensors to measure a person's heartbeat and will visualize and store the heartbeat data in an Android smartphone

2

Classification

The system could also provide an **automatic classification of the person's cardiovascular health**. In addition to that, the system also sends the person's data to a doctor.

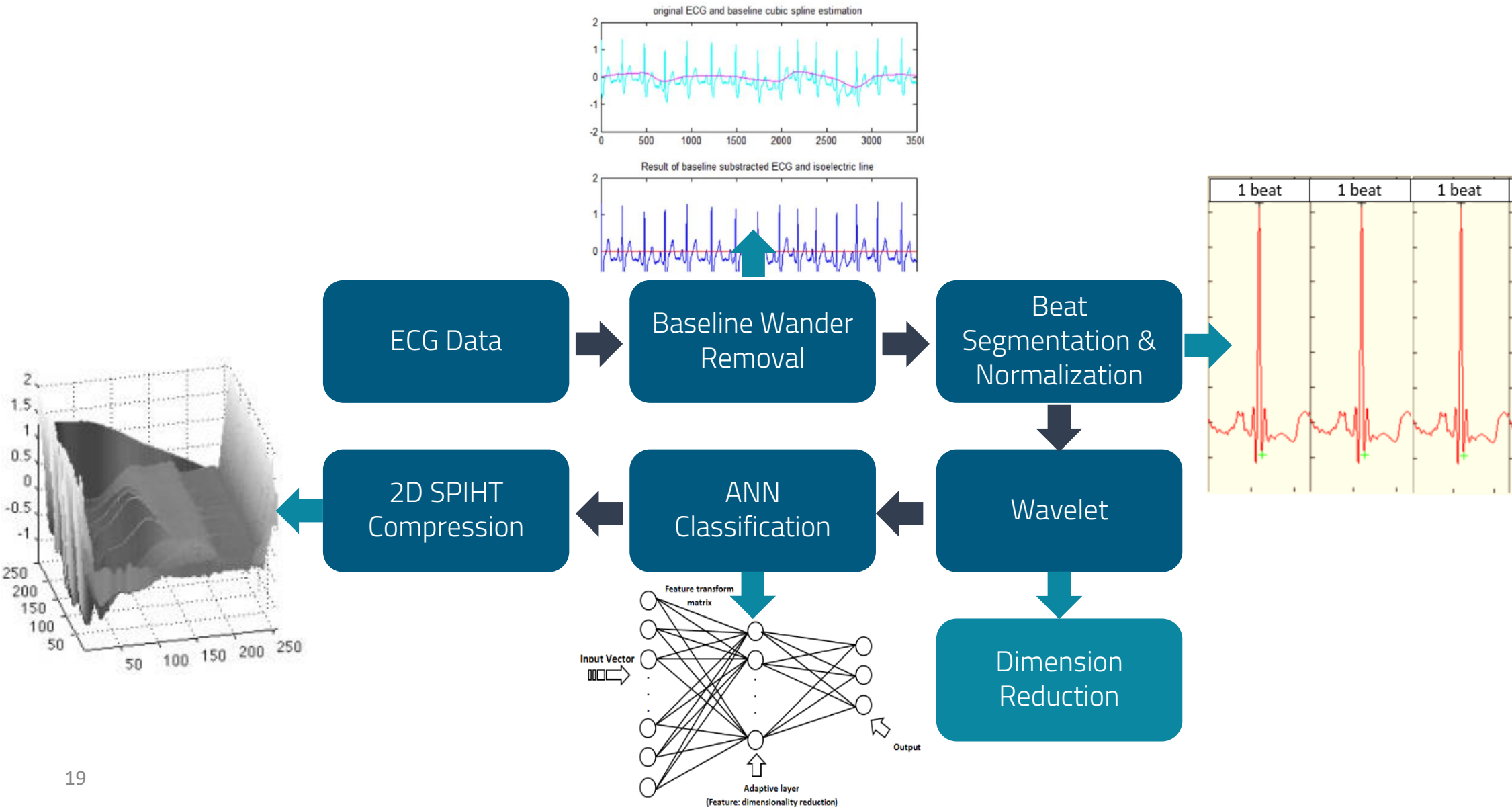
3

Transmission

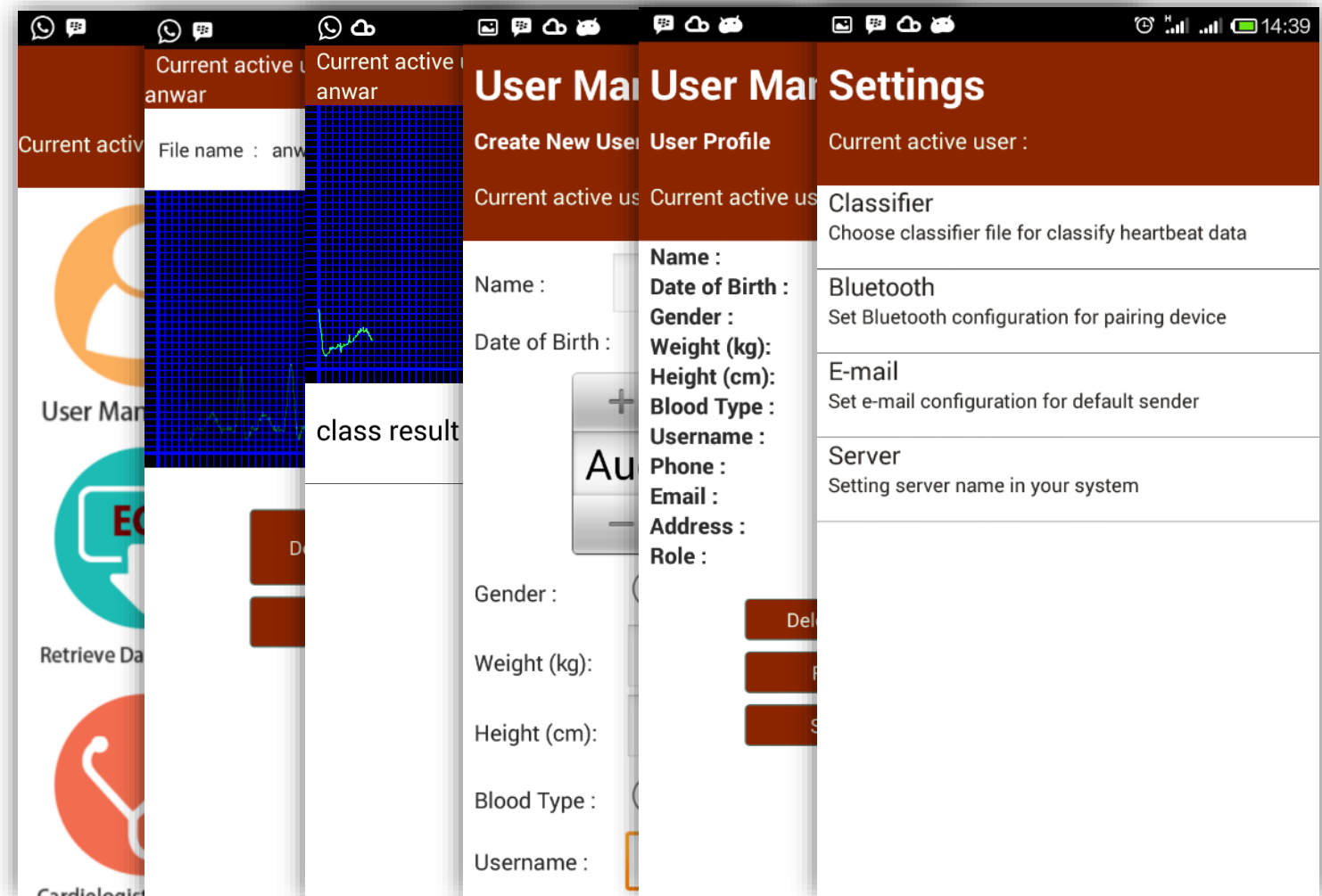
Developed a method for ECG signal compression to be transmitted via cellular signal



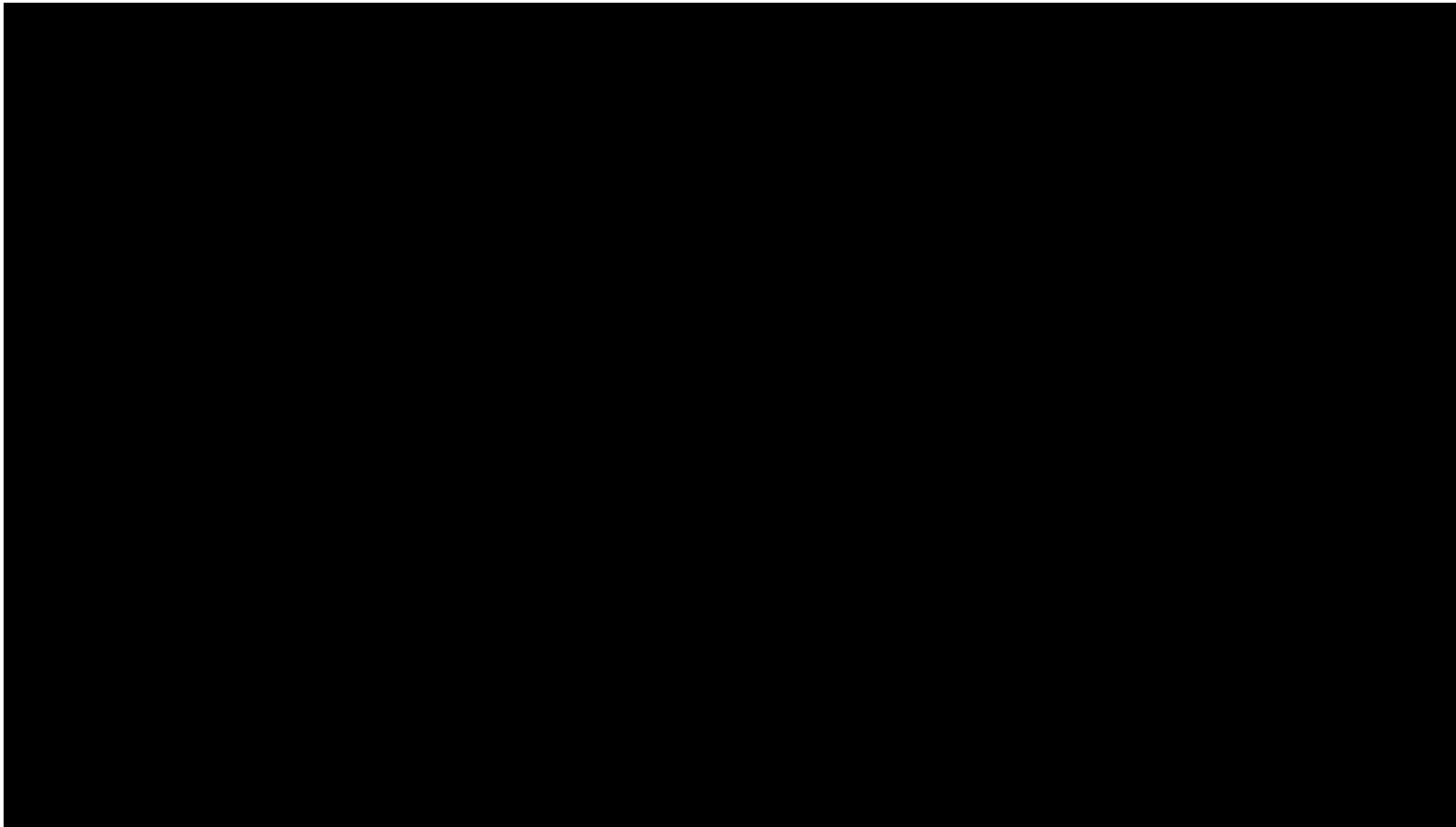
AI- Based Heart Beat Classification



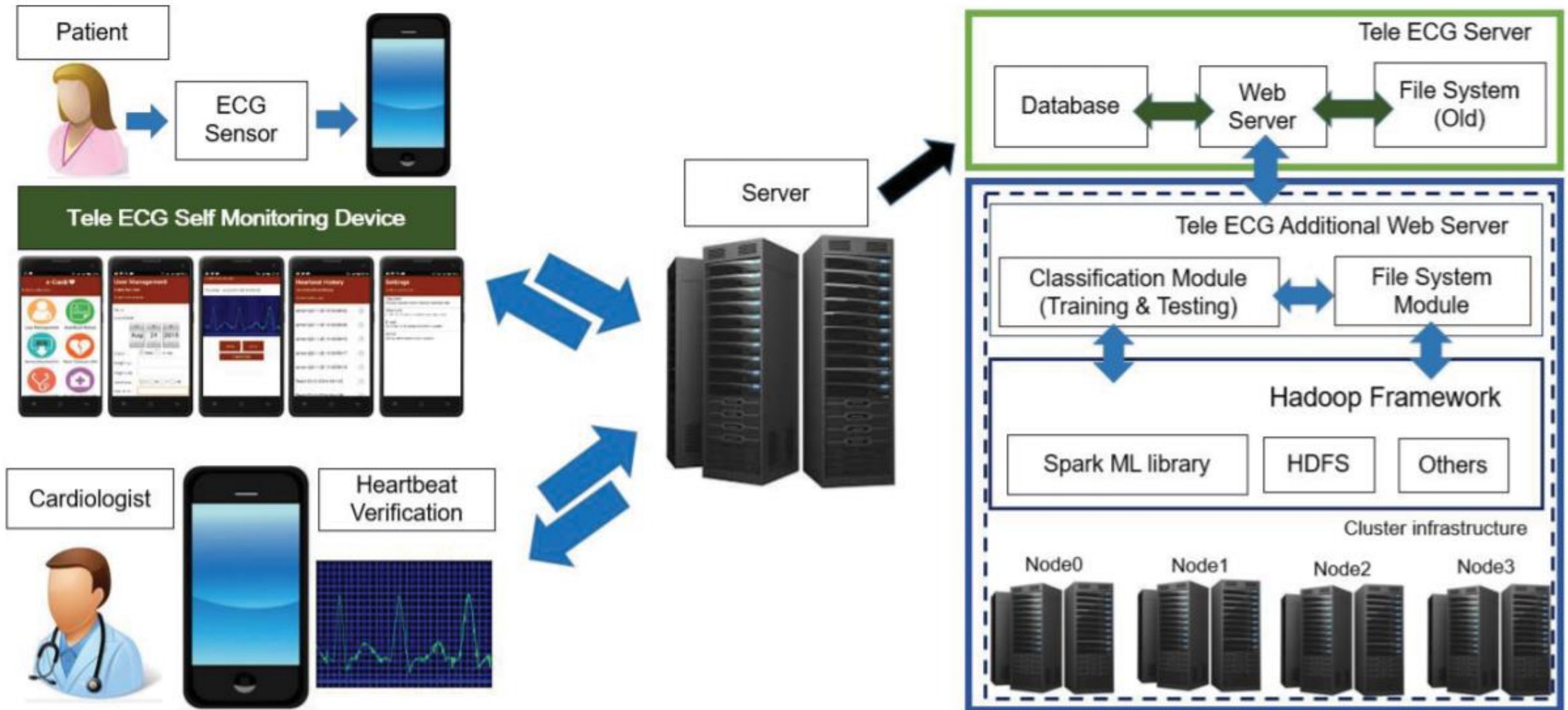
Heart Beat Classification - System



Tele-ECG Demo (Video)



Implement AI Application in Big Data Framework



Patient in house or every where
Takes vital signs and ECG surveys
tailored to the **USG Self
Monitoring Device**



Infrastructures
With a special algorithm, the server process clinical
data into information. Managed with two directional
data communication with electronic health records.



Doctor
Where ever and When Ever can
annotate, discussion and give a
handling



Nurse
Clinical software allows for
earlier intervention patients.



Ultrasonography Fetal Monitoring

USG System

In our Tele-USG, a complete system have been developed. The main feature of our system is detecting body parts of the fetus to reduce the risk of death in pregnancy by monitoring the growth rate of the fetus.

1

**Fetal Organ
Segmentation**

2

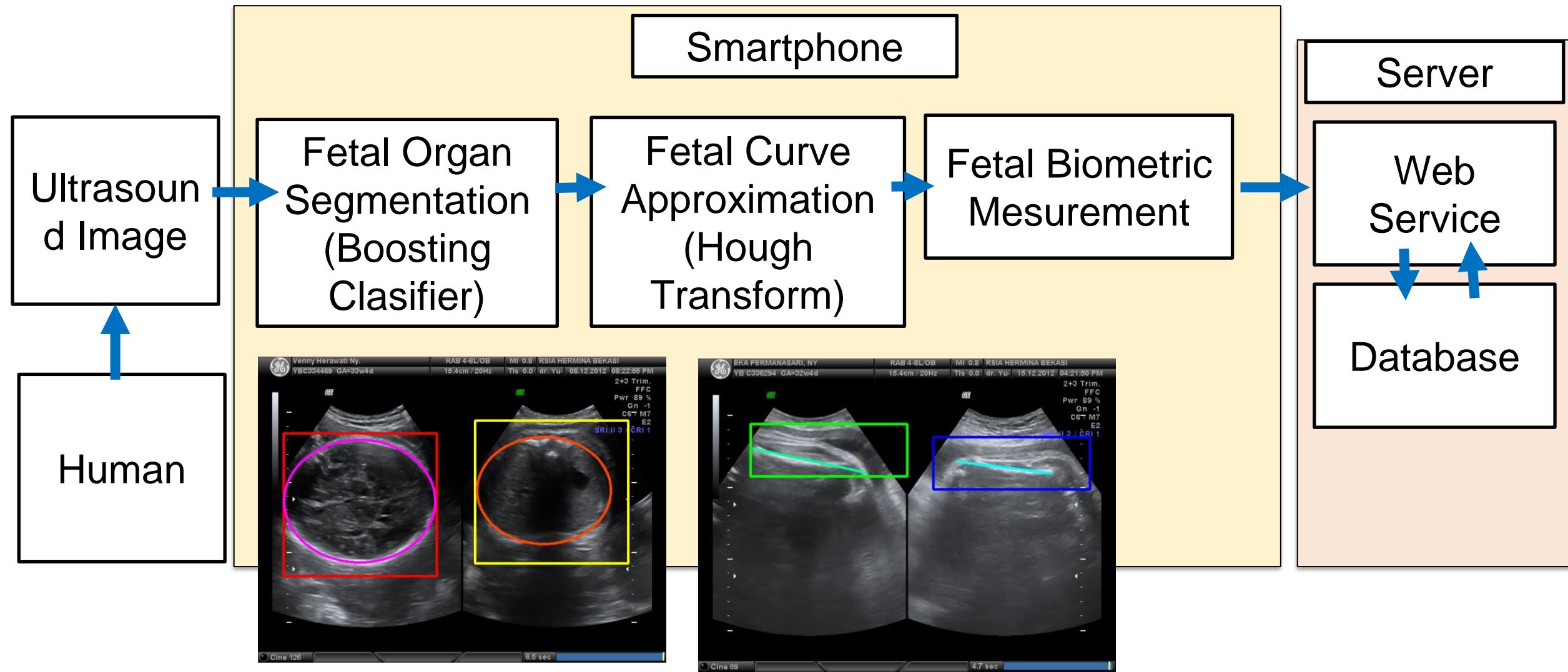
**Fetal Organ
Approximation**

3

**Fetal Biometric
Measurement**

Product Video

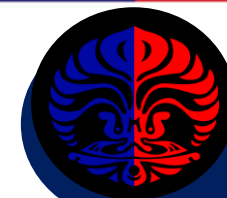
AI-Based Fetal Detection



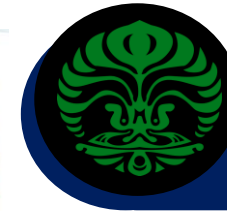
Fetal Detection - System



Telehealth Consortium



Fasilkom UI



FK UI



Kemkominfo



Solusi247



Samsung R&D Institute

TKT 9

TKT 8

TKT 7

TKT 6

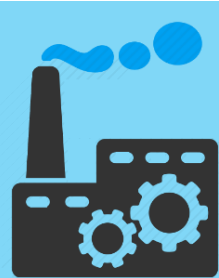
TKT 5

TKT 4

TKT 3

TKT 2

TKT 1



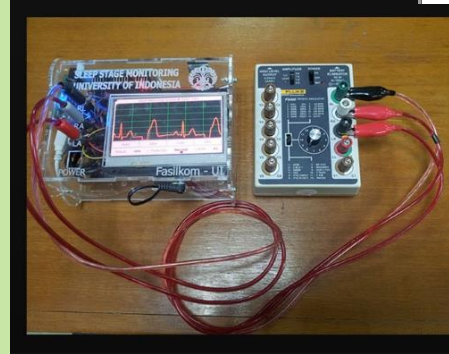
Product



Prototype & Copyright



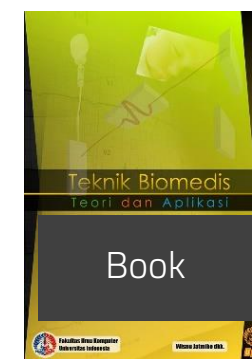
Science Article



International Journal



International Conference



Book

An abstract digital graphic on a dark background. It features a network of orange lines and dots, some of which form a low-poly map of the African continent. In the background, there are faint, circular patterns of white dots and various geometric shapes like hexagons and squares, some with arrows indicating flow or direction. The overall aesthetic is technological and futuristic.

END OF PRESENTATION