

Towards Efficient and Safe ML in Real-Time Cyber-Physical Systems

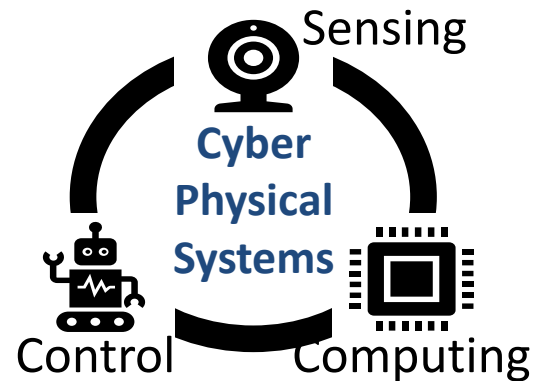
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ML-RT-Agenda 2025 @Brussels

Cyber-Physical Systems (CPS)

- Engineered systems that are built from, and depend upon, the seamless integration of computation and physical components.
- Real-Time Systems
 - *both logical and temporal* Correctness
 - **Predictability**: Provably guaranteed responses within strict time constraints (deadlines)



Efficient Computing & Automation

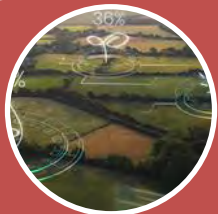
Real-Time Scheduling & Control

Wireless Networking

Wireless Sensing

Formal Methods

Constrain-Aware Security & Privacy



Agriculture



Environmental



Smart City



Healthcare

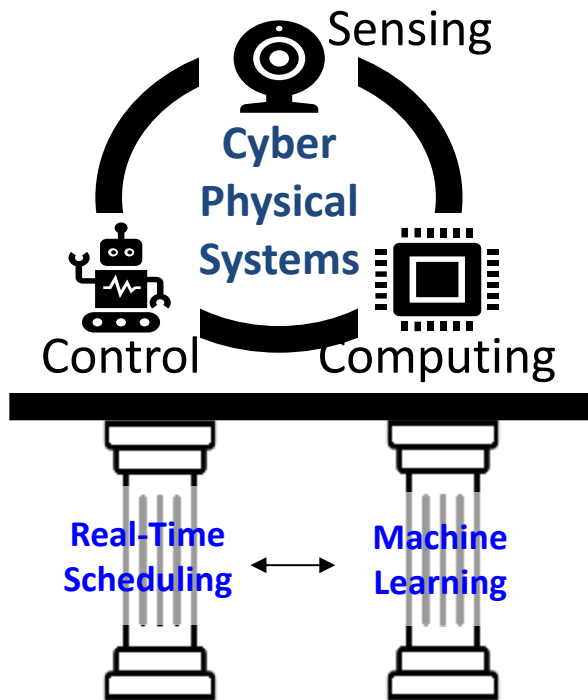


Robotics &
Manufacturing



Autonomous
Driving

Towards Efficient and Safe ML in Real-Time CPS



- Key **Challenges** in RT-CPS Design and Analysis
 - System Uncertainty & Reliability/Safety
Application 1: F1-Tenth Autonomous Racing
 - Resource and Energy Efficiency
Application 2: Cardiovascular Monitoring
 - Scalability and Curse of Dimensionality
Application 3: Gait Kinematics Estimation

• Mixed Criticality Real-Time OS

[RTCSA'21] Reserving Processors by Precise Scheduling of Mixed-Criticality Tasks.

[RTNS'21] Precise Scheduling of Mixed-Criticality Tasks on Varying-Speed Multiprocessors.

[ICCAD'20] F2VD: Fluid Rates to VDs for Precise MC Scheduling on a Varying-Speed Processor.

[RTSS'18] Precise Scheduling of Mixed-Criticality Tasks by Varying Processor Speed.

• CPU-GPU Co-Scheduling

[RTSS'19, TIME'21] MC scheduling of Gang tasks

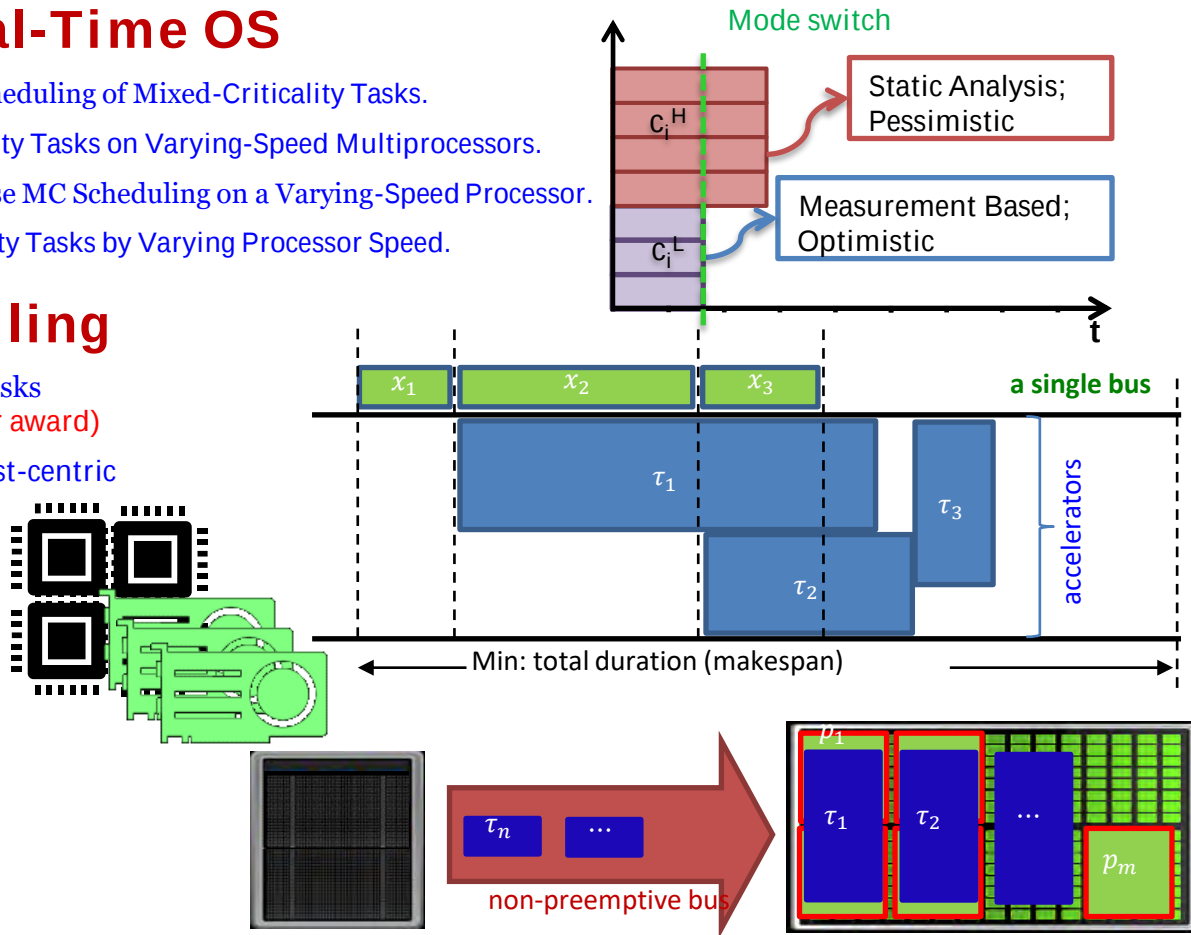
(Outstanding paper award, Best student paper award)

[RTAS 2020] Real-time scheduling upon a host-centric acceleration architecture with data offloading

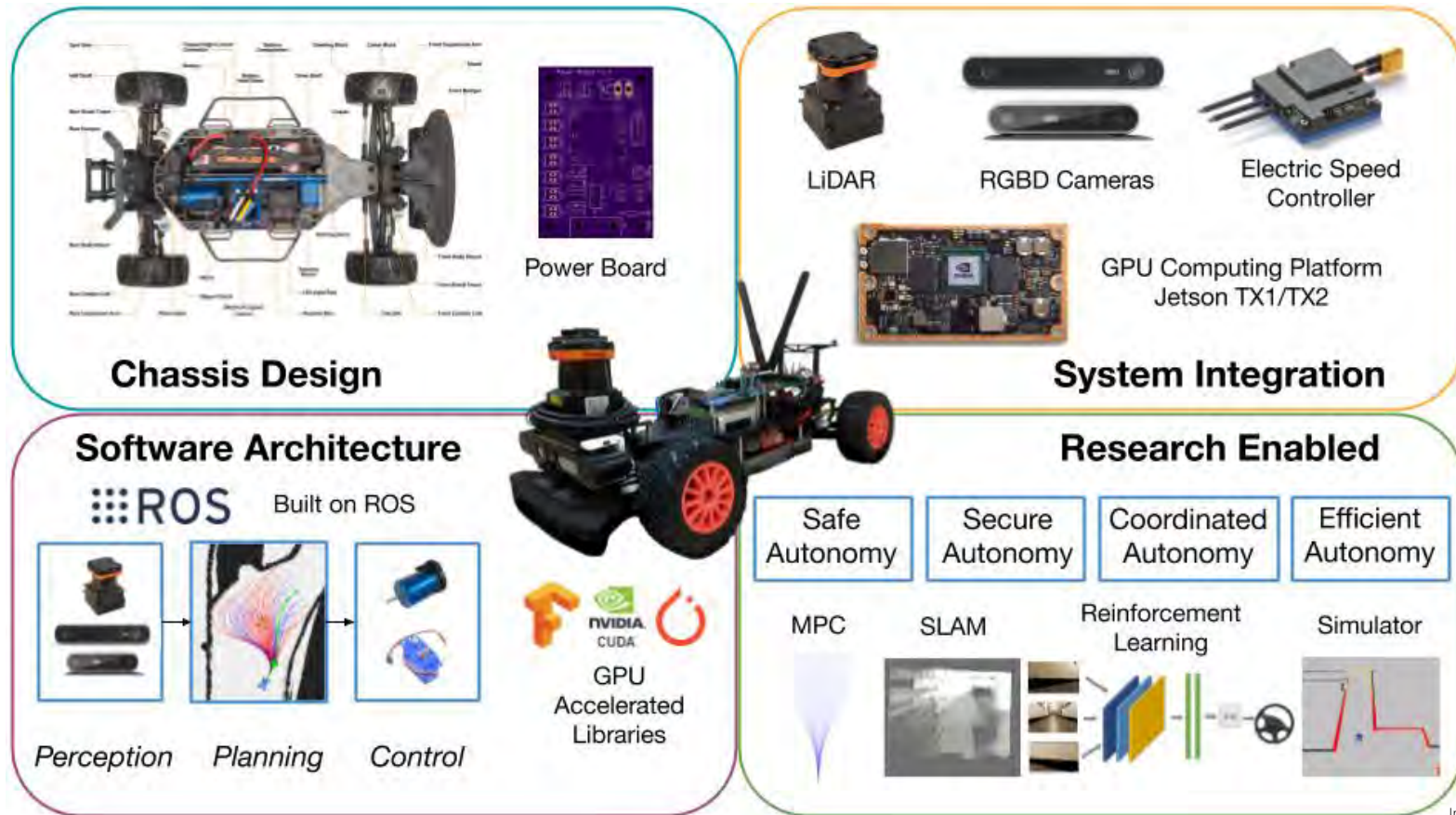
• IDK-Cascades

[RTS 2024] Scheduling IDK classifiers with arbitrary dependences to minimize the expected time to successful classification

[ECRTS'25] Optimizing IDK Cascades for Time-Series Input Streams



• Application 1: F1-Tenth Autonomous Racing



ROS2 3-layer architecture: Software/Run-Time/RTOS

[RTAS'25] Physics-Informed MC Scheduling for F1Tenth Cars with Preemptable ROS 2 Executors

[HSCC'25] SOTERIA: A Formal Digital-Twin Framework for Safety-Assurance of Latency-Aware Safety-Critical CPS

[TCAD 2024] Dynamic Priority Scheduling of Multithreaded ROS 2 Executor With Shared Resources

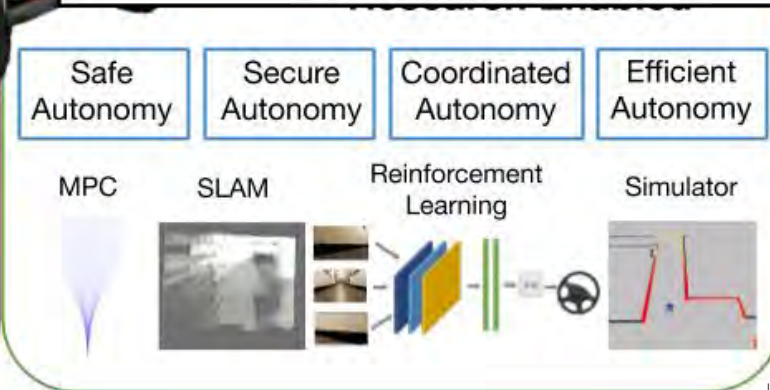
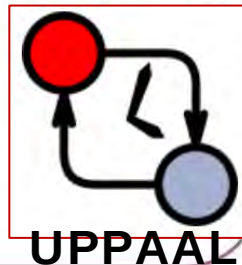
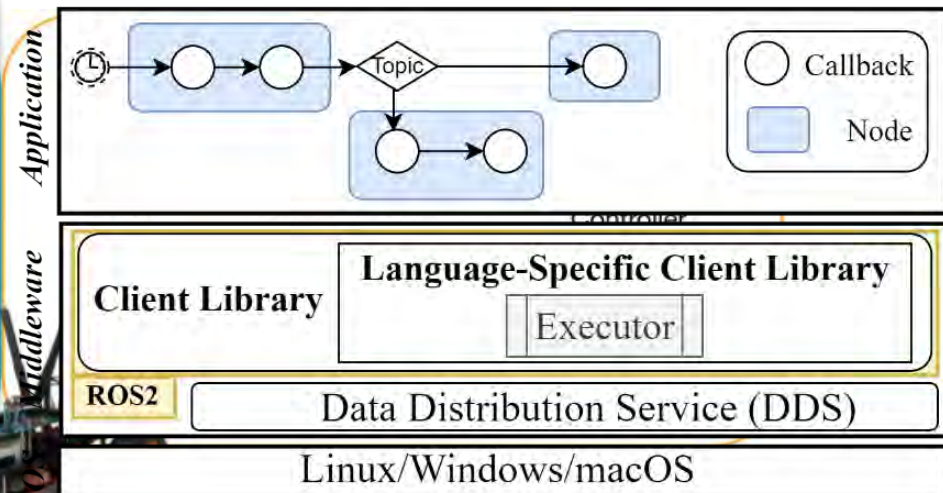
[RTSS'23] SEAM: An Optimal Message Synchronizer in ROS with Well-Bounded Time Disparity (Best Paper Award)

[RTAS'23] Real-Time Scheduling of Autonomous Driving System with Guaranteed Timing Correctness

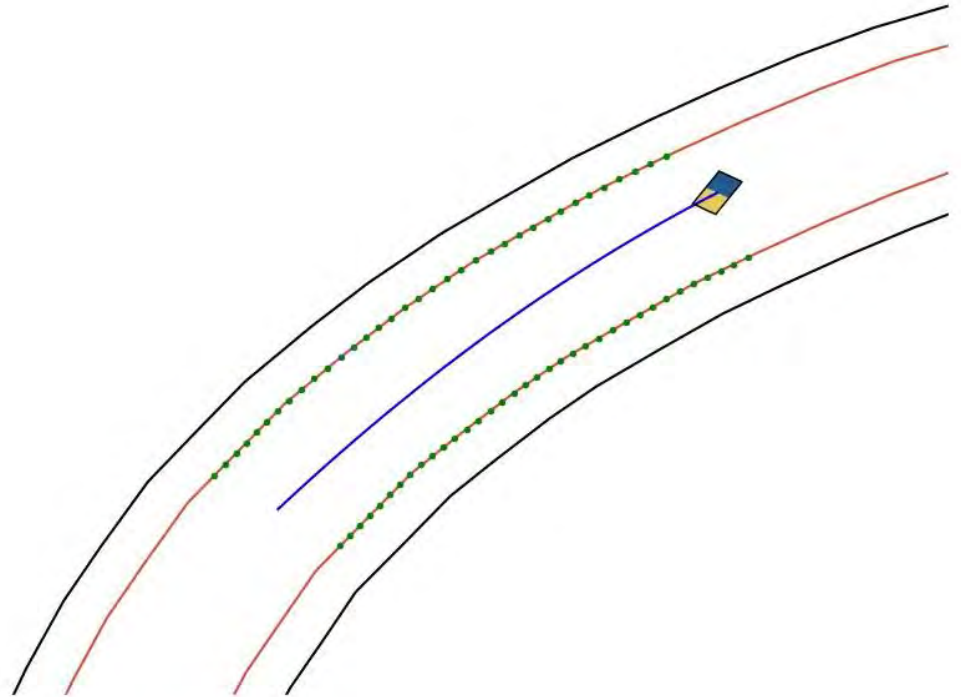
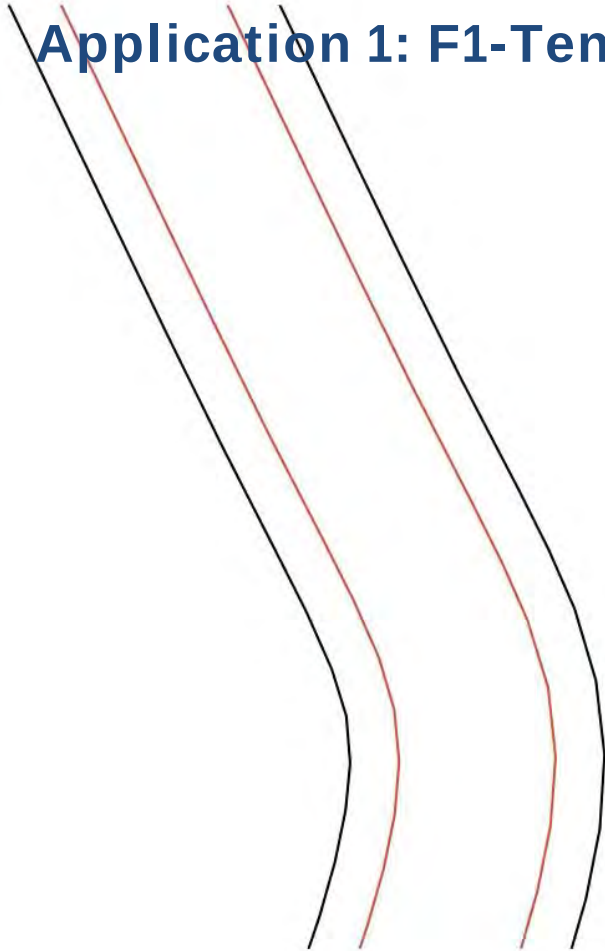
[DAC'22] Response-Time Analysis for D'line-Based Scheduling of ROS2

[RTSS'21-Industry] Toward Real-Time Guaranteed Scheduling for Autonomous Driving Systems (Best Solution Award)

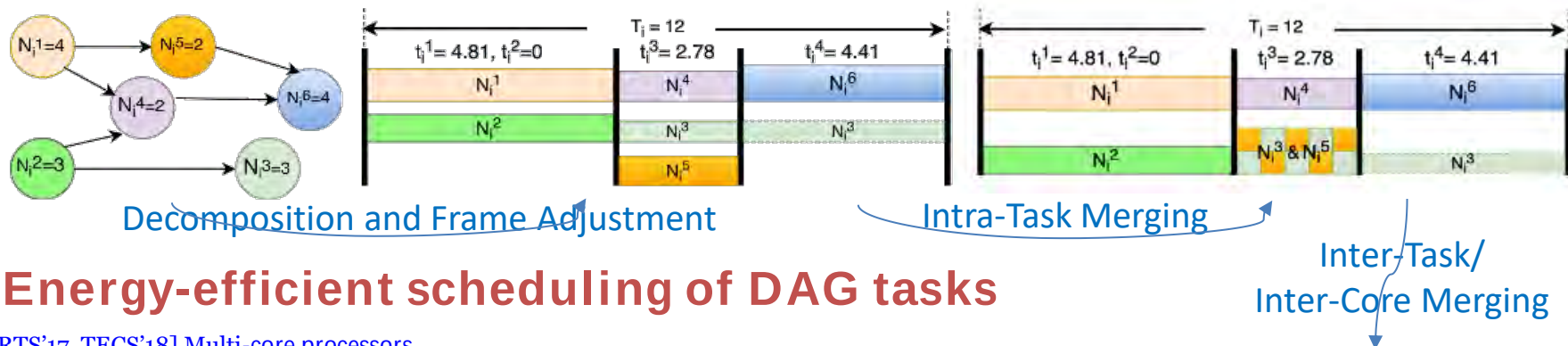
Autonomous Racing



- **Application 1: F1-Tenth Autonomous Racing**



DAG/Graph-Based Task



Energy-efficient scheduling of DAG tasks

[ECRTS'17, TECS'18] Multi-core processors

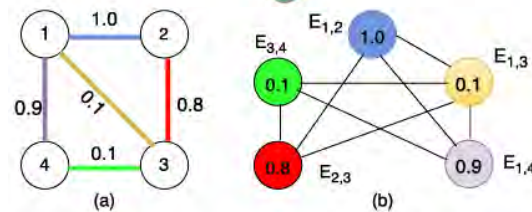
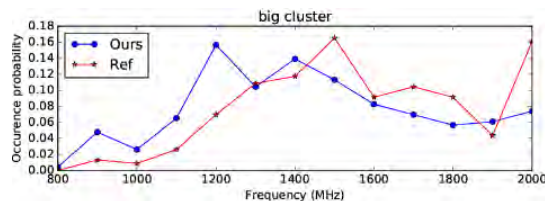
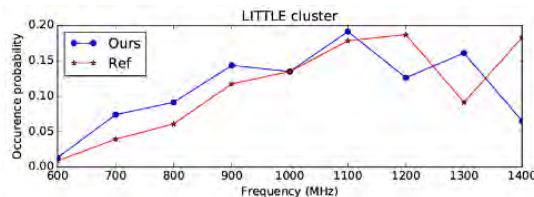
[RTAS'19, TPDS'20] Clustered multi-core processors

Response time and feasibility analysis

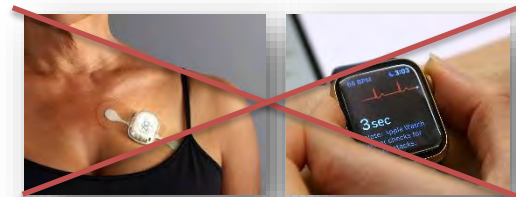
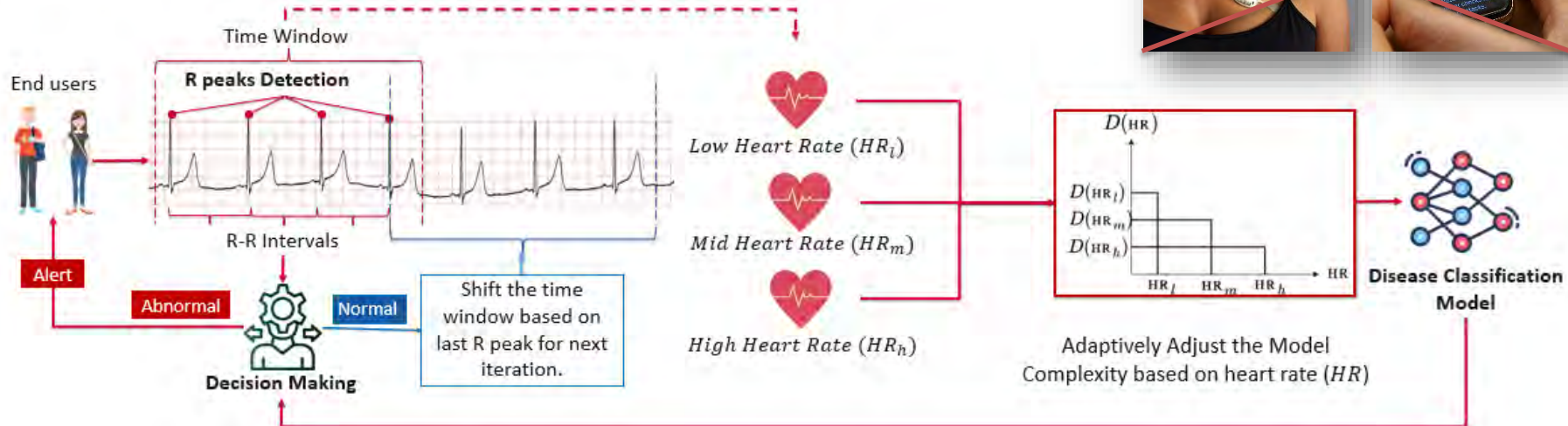
[RTSS'21] Calculating Worst-Case Response Time Bound for OpenMP Programs with Loop Structures.

[EMSOFT'20] Efficient Feasibility Analysis for Graph-based Real-Time Task Systems (Best paper award)

[DAC'20] On Computing Exact WCRT for DAG Task



• Application 2: Real-time Cardiovascular Disease Detection on Embedded Devices



RQ1: How to balance the accuracy and inference speed, while preserving the memory usage?

[RTCSA'25] Adaptive Model Selection for Real-Time Cardiovascular Disease Detection on Embedded Systems

[ISNN'24] CardiacRT-NN: Real-Time Detection of Cardiovascular Disease Using Self-attention CNN-LSTM for Embedded Systems

[IPSN'24] Real-Time cardiovascular disease detection via abnormal electrocardiogram cycles on embedded systems

[ICCPs'15] Uniprocessor EDF scheduling of AVR task systems



Heart Rate:
Low



Heart Rate:
Mid

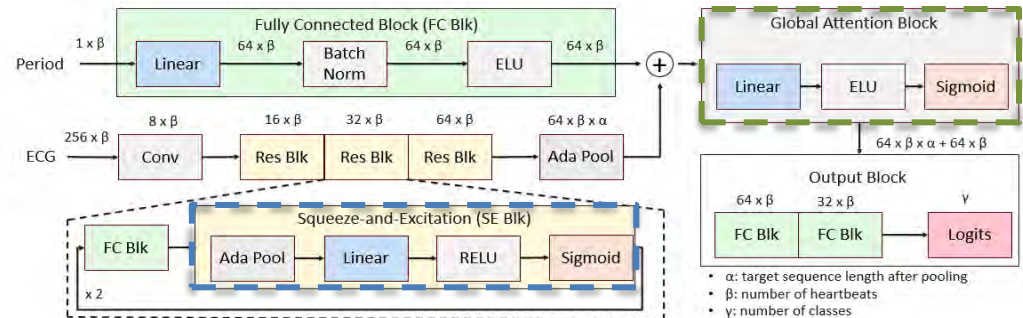
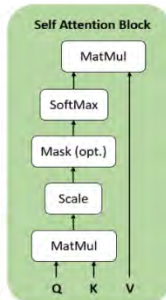
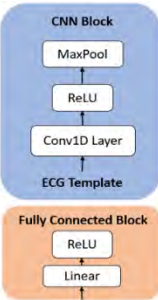
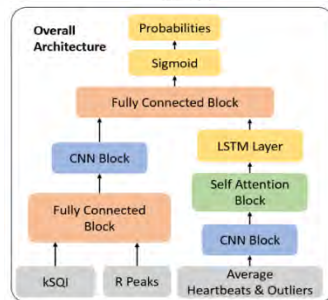
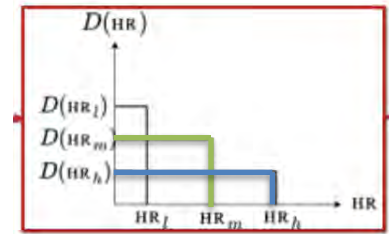
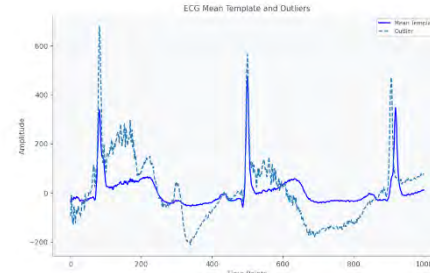
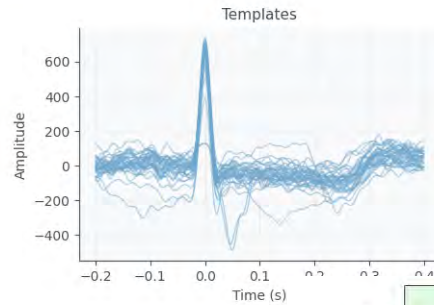
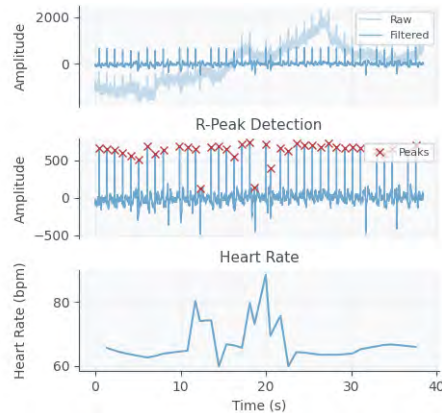


Heart Rate:
High

Application 2: Real-time Cardiovascular Disease Detection

RQ2: How to design highly accurate DL models for abnormality detection?

RQ3: How to design a framework to handle varying heart rate conditions?



Self-Attention CNN-LSTM

- α : target sequence length after pooling
- β : number of heartbeats
- γ : number of classes

High Dimension and Low Sample Size (HDLSS) Settings

When the number of training samples is less than the dimension size of training sample, it can be treated as HDLSS settings.

- **LDA: inverse covariance matrix ???**

(sparse regularization/ de-biased)

[ICDM'17] AWDA: An Adaptive Wishart Discriminant Analysis

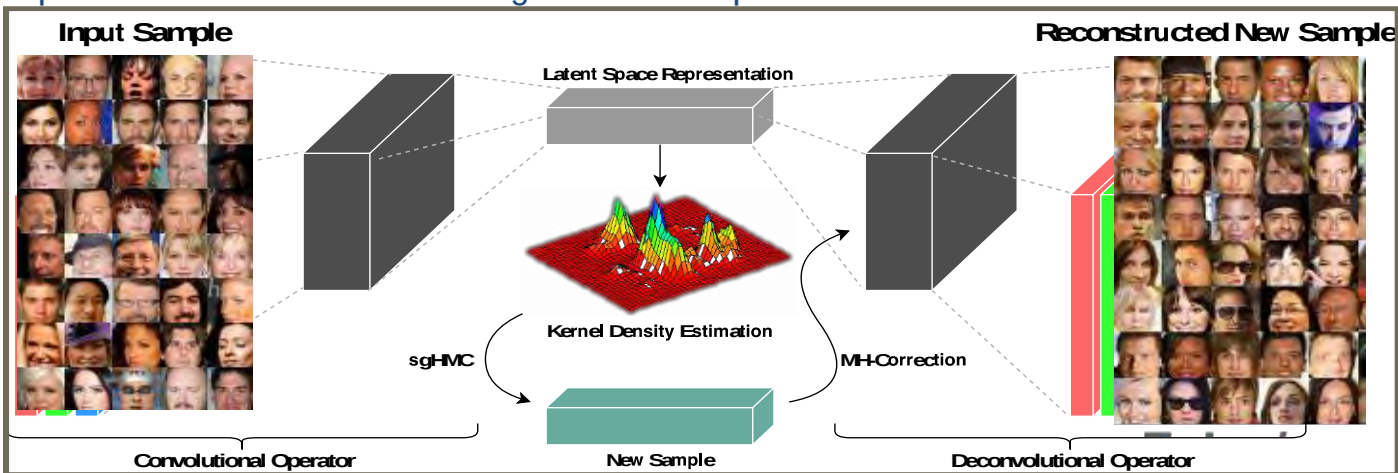
[IJCAI'18] De-Biasing Covariance-Regularized LDA

[TNNLS'20] DBSDA: Lowering the Bound of Misclassification Rate for Sparse LDA via Model Debiasing

- **Dataset Generation**

- Latent space explore for high quality sample generation
- Estimated kernel density distribution
- Explainable solution with strong control on space search

[AAAI'19] SpHMC: Spectral Hamiltonian Monte Carlo

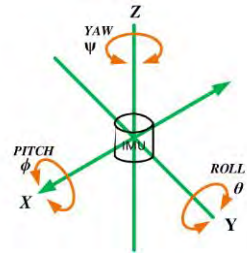


Application 3: Gait Kinematics Estimation

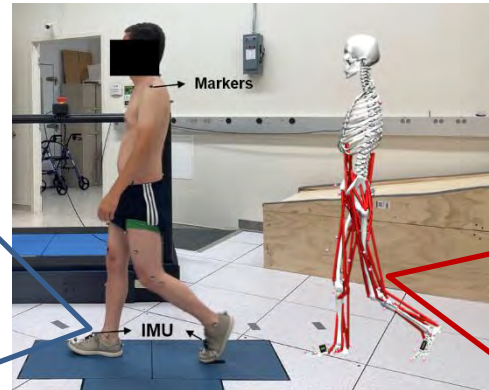
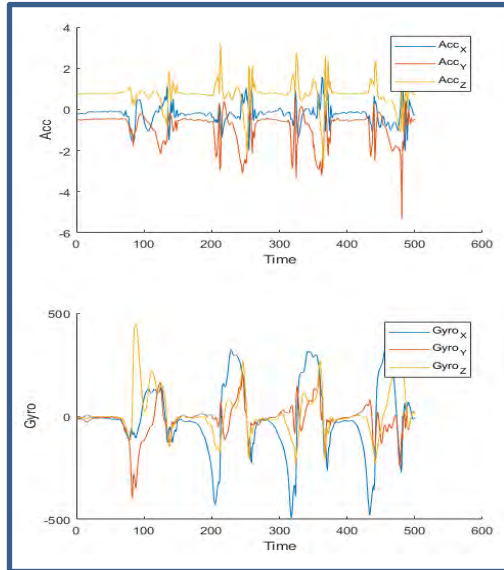
- Kinematics/joint angle
 - Disease progression
 - Kinesiological health



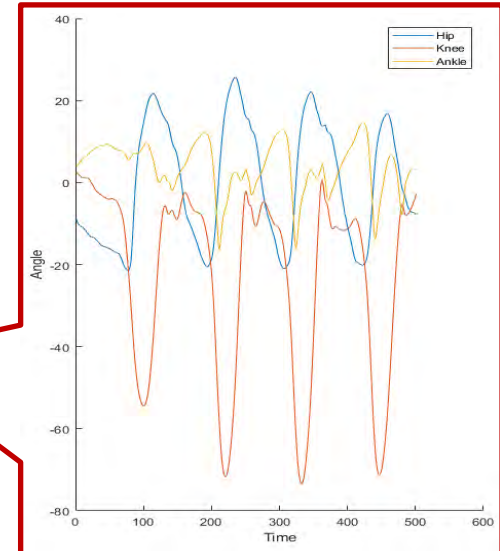
3 Axes
Accelerations



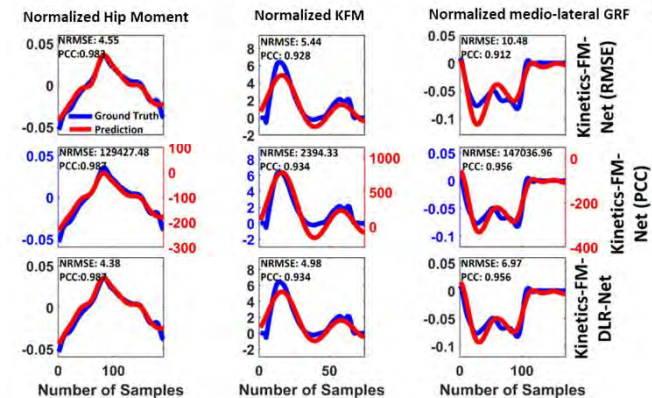
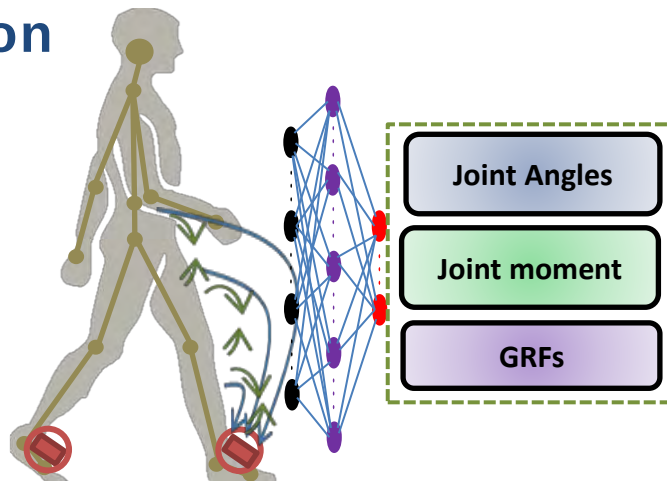
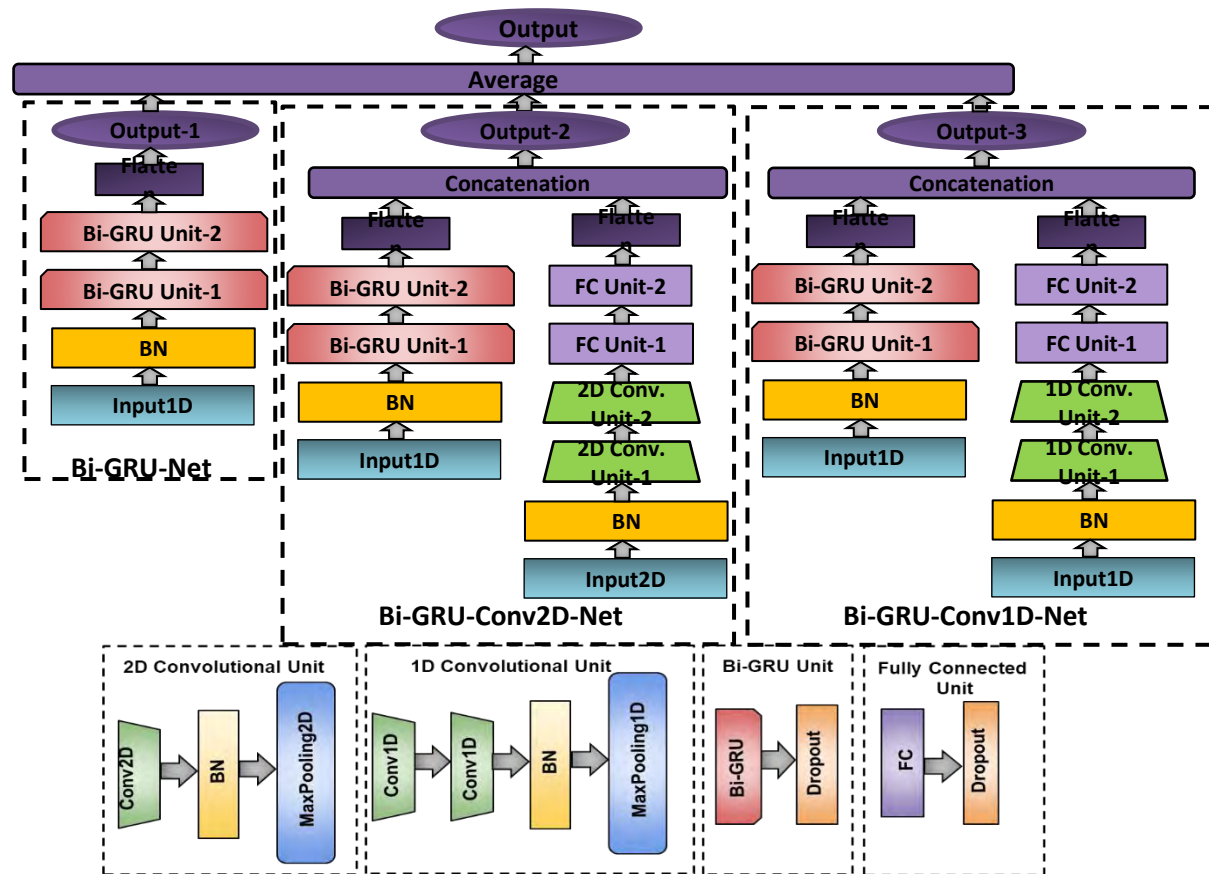
3 Axes
Angular Velocity



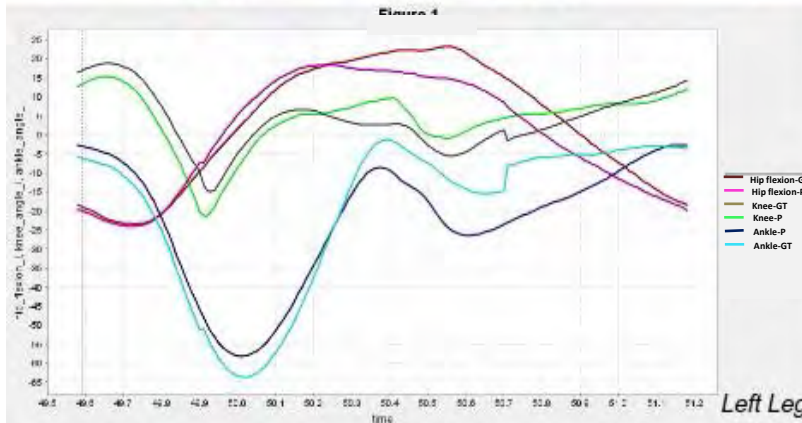
~ \$10
Compact Size
Light Weight



Application 3: Gait Kinematics Estimation



• Application 3: Gait Kinematics Estimation



[HICSS'24] Predicting Lower Extremity Joint Kinematics Using Multi-Modal Data in the Lab and Outdoor Environment

[JBHI2023] Wearable Motion Capture: Reconstructing and Predicting 3D Human Poses from Wearable Sensors

[JBHI2023] Estimation of Lower Extremity Joint Moments and 3D Ground Reaction Forces Using IMU Sensors in Multiple Walking Conditions: A Deep Learning approach

[JBHI2023] DeepBBWAE-Net: A CNN-RNN Based Deep Super Learner for Estimating Lower Extremity Sagittal Plane Joint Kinematics Using Shoe-Mounted IMU Sensors in Daily Living

[Chase'22] Estimation of Hip, Knee, and Ankle Joint Moment Using a Single IMU Sensor on Foot Via Deep Learning

Thank You!

PhD Students and Postdocs

Dr. Jiang Bian, Baidu Research
Prof. Ashik Ahmed Bhuiyan, West Chester U
Prof. Federico Reghenzani, Politecnico di Milano
Dr. Sudharsan Vaidhun, Cummins
Dr. Seyed Iman Taheri, University of Memphis
Dr. Ashkan Farhangi
Dr. Md Sanzid Bin Hossain
Abdullah Al Arafat
Kurt Wilson
Yixin (Eason) Li
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