



# BALANCE

Behavioral and Ambient Learning Across Networked Collaborative Edges

An overview of Methodology and  
Results of Neural Networks for Sleep  
Quality Prediction

# *Agenda*

- 1. Problem Overview and Task Definition**
- 2. Ex Ante Literature Review**
- 3. Framework Adopted**
  - 3.1. Data Engineering**
  - 3.2. Network Architecture**
- 4. Results**
- 5. Ex Post Literature Review, Considerations and Future Works**



# *Problem Overview and Task Definition*

Sleep quality has been proven through extensive research to be a key element academic performance [1][2] ...



[1] Lorenzo Tonetti, Marco Fabbri, Marco Filardi, Monica Martoni, Vincenzo Natale, Effects of sleep timing, sleep quality and sleep duration on school achievement in adolescents, ISSN 1389-9457,

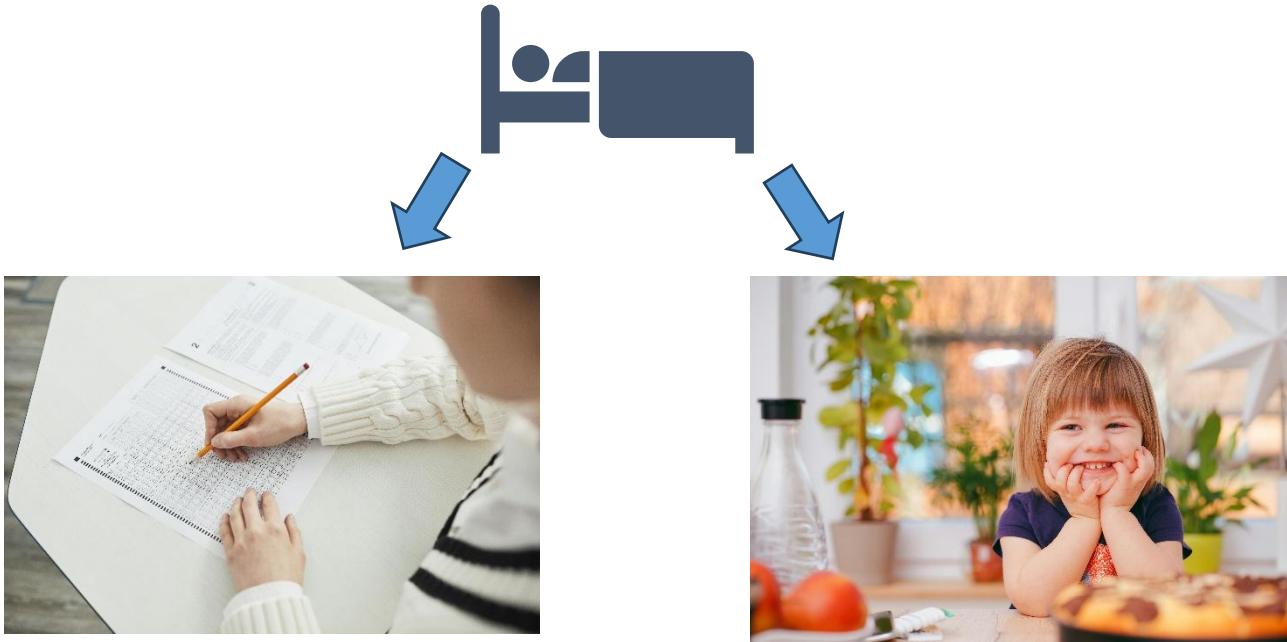
[2] Julia F. Dewald, Anne M. Meijer, Frans J. Oort, Gerard A. Kerkhof, Susan M. Bögels, The influence of sleep quality, sleep duration and sleepiness on school performance in children and adolescents: A meta-analytic review, ISSN 1087-0792

[3] Shin Ji-eun , Kim Jung Ki , How a Good Sleep Predicts Life Satisfaction: The Role of Zero-Sum Beliefs About Happiness, ISSN 1664-1078



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... But also on Happiness and Optimism towards the future! [3]

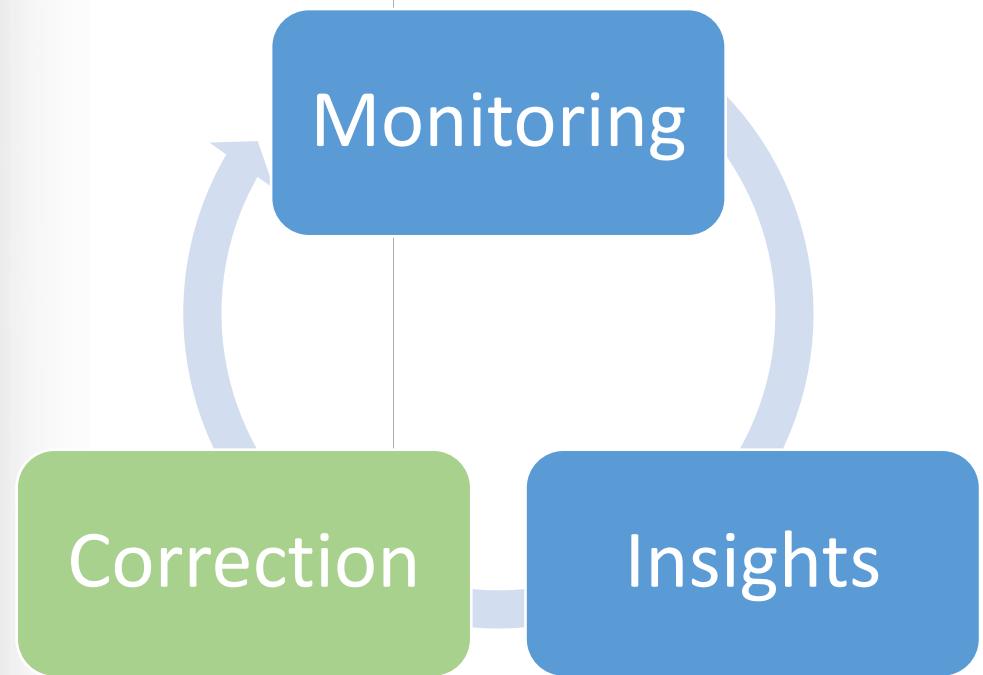
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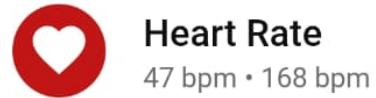
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49:39 • 0,51 km



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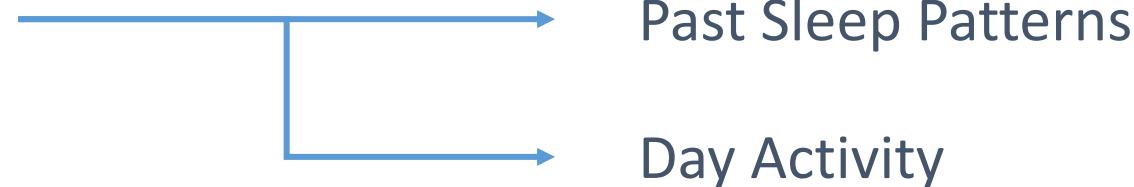
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To do so, we use a **Federated Approach**, which allows us to keep the data from different users **separated**, respecting privacy of the users



## Relevant Features

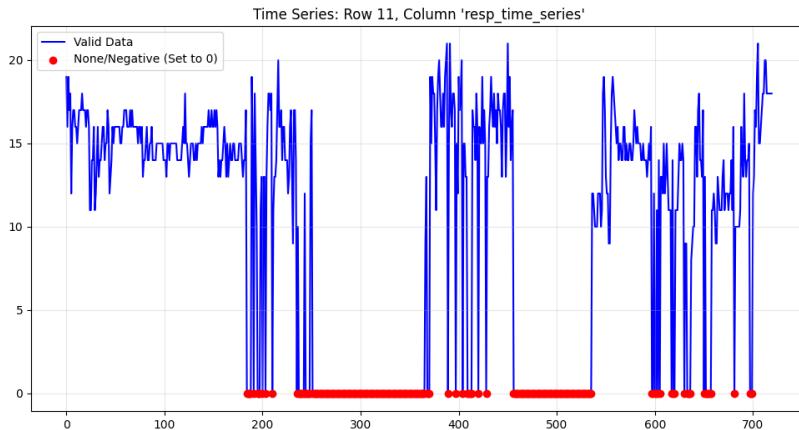


## Chosen Architectures



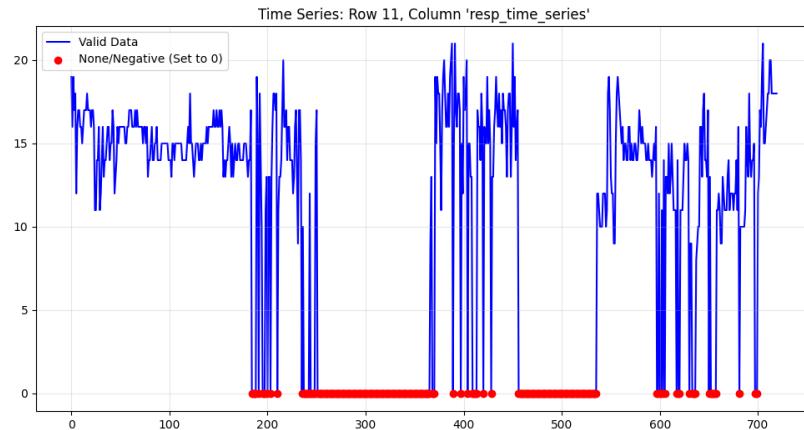
Salgaonkar, Sujal & Ghule, Kishori. (2025). Predicting Sleep Quality Using 24-Hour Physical Activity Data from Wearable's: A Fitness Tracking Approach. International Journal of Scientific Engineering and Research. 13. 27-32. 10.70729/SE25115123316.  
Sathyaranayana A, Joty S, Fernandez-Luque L, Ofli F, Srivastava J, Elmagarmid A, Arora T, Taheri S, Sleep Quality Prediction From Wearable Data Using Deep Learning, JMIR Mhealth Uhealth 2016;4(4):e125, DOI: 10.2196/mhealth.6562  
Arora, A., Chakraborty, P. & Bhatia, M.P.S. Analysis of Data from Wearable Sensors for Sleep Quality Estimation and Prediction Using Deep Learning . Arab J Sci Eng 45, 10793–10812 (2020). <https://doi.org/10.1007/s13369-020-04877-w>

# Framework Adopted – Data Engineering

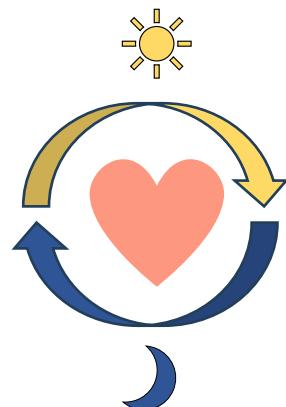


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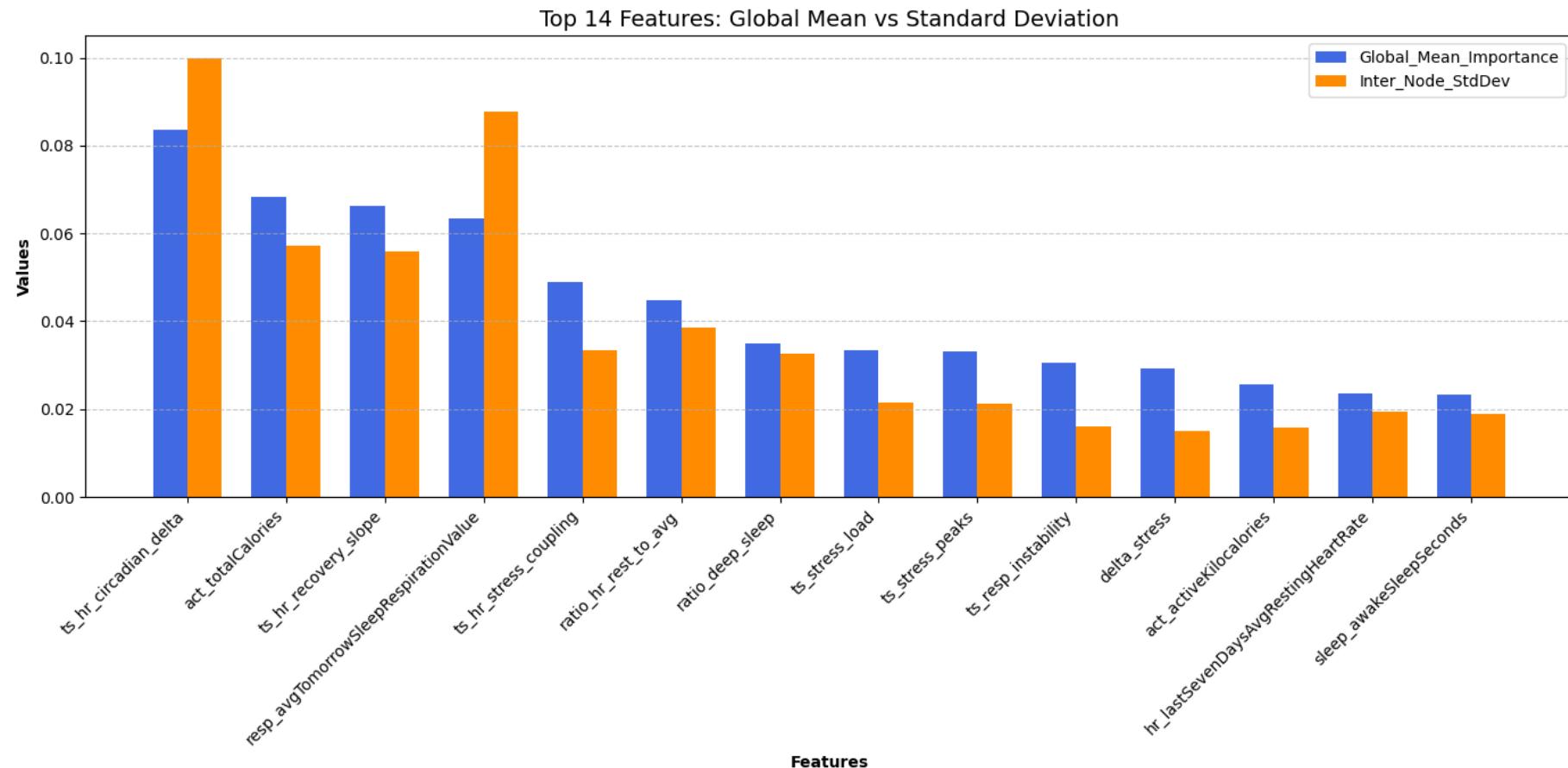


$$\frac{\text{Active KCal}}{\text{Total KCal}}$$

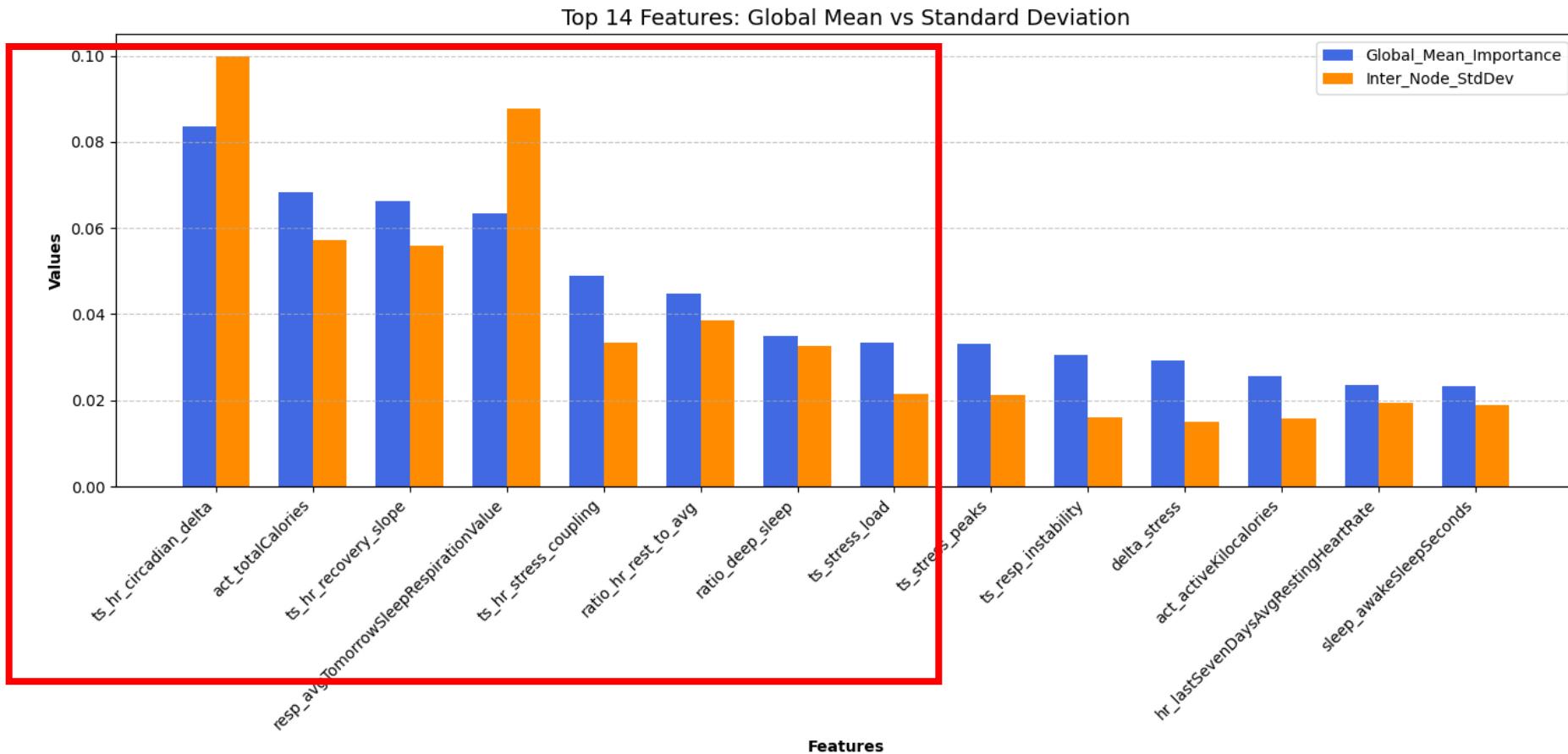
## Addition of

- Circadian Features
- Early Sleep Quality
- LogKCal
- Ratios of Existing features

## Random Forest – based Feature Extraction

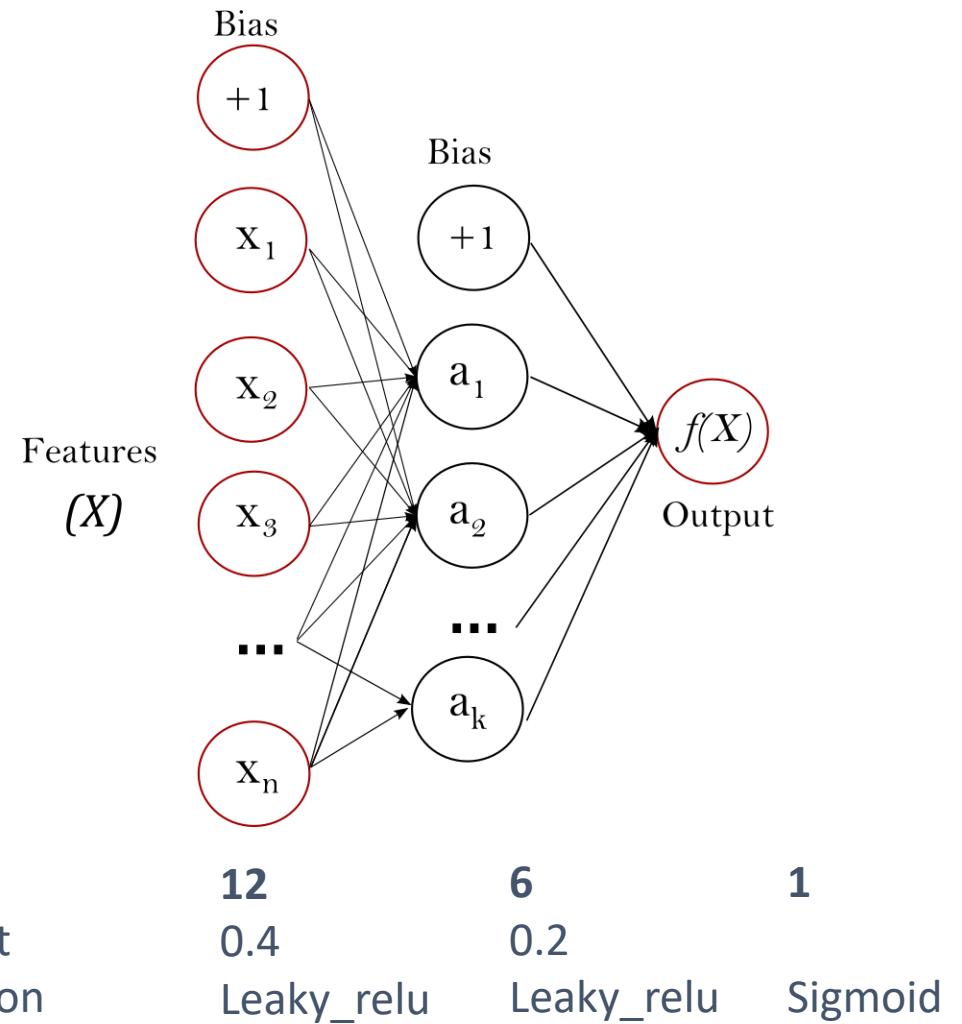


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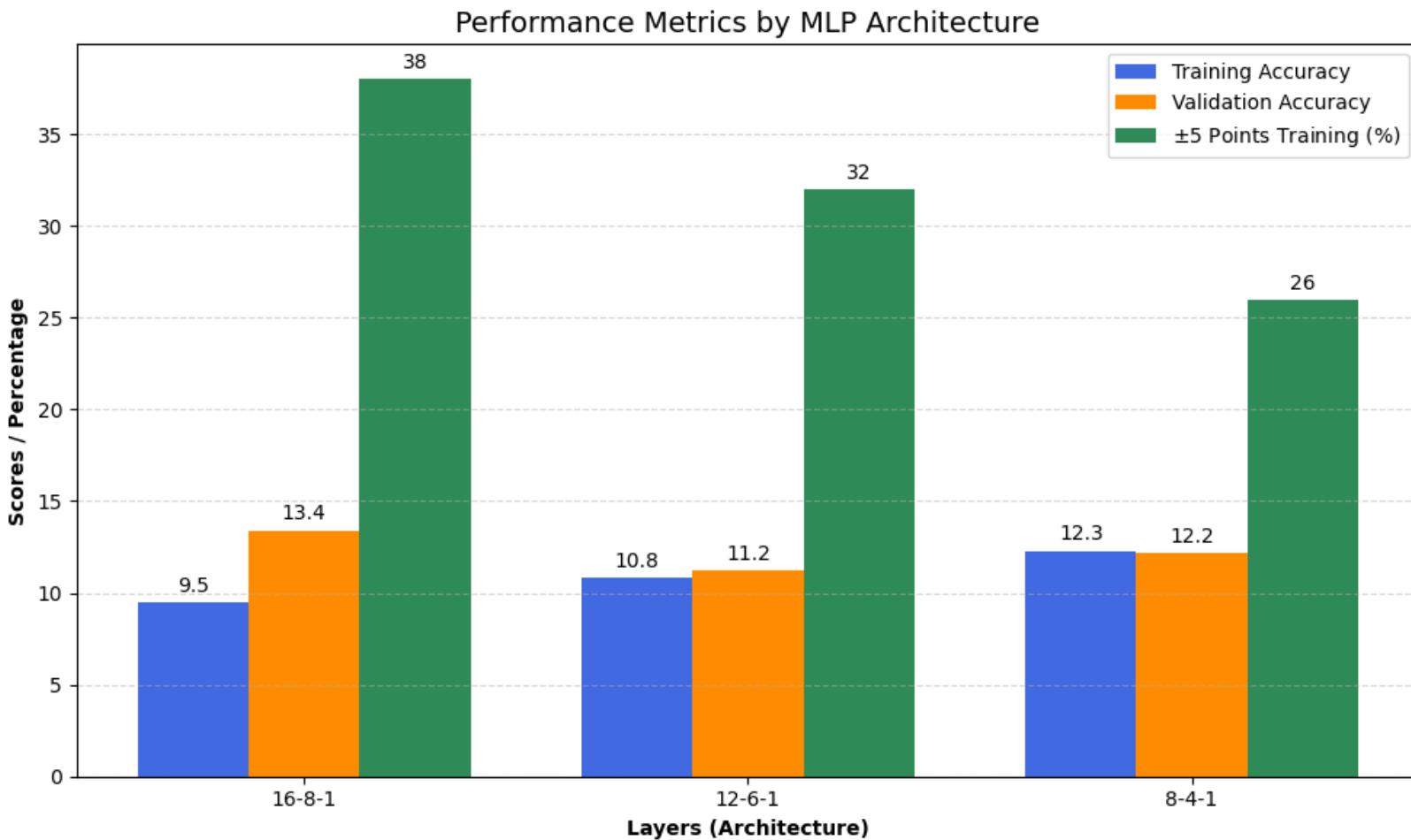


# Framework Adopted – Network Architecture

Parameter	Value
Regularization	L2(0.001)
LossFunction	MSE
Batch Size	16
Local Epochs	8
Round	35
Averaging	FedProx



# Results



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**Variance of Prediction consistently doesn't reflect variance of label in the datasets**

**Local / Federated Discrimination to match local / global distribution**



# Thank You!



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