

SLEEP 2024

HOUSTON, TX
JUNE 1-5

Improved Sleep Apnea
Screening After Stroke
Using Multimodal
Wearable Sensors and
Machine Learning

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A JOINT MEETING

AASM American Academy of
SLEEP MEDICINE™

S Sleep Research Society®
Advancing Sleep & Circadian Science

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Learning Objectives

Upon completion of this activity, participants should be able to:

- Understand the need to monitor sleep early in patients with stroke
- Identify the importance of patient acceptance in sleep apnea screening
- Describe the advantages of machine learning and wearable devices for sleep apnea screening

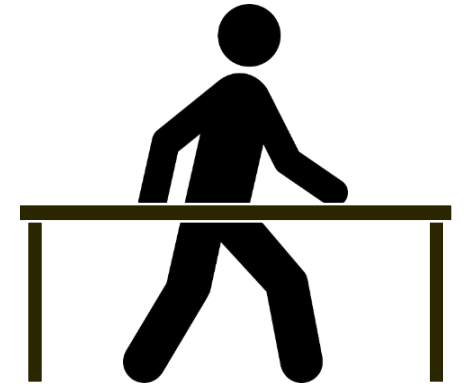
Background

Sleep is an essential factor in the stroke recovery process

Sleep facilitates:

- neural plasticity, learning, and memory consolidation.¹⁻³

But, **50-70%** of stroke survivors suffer from sleep apnea.⁴

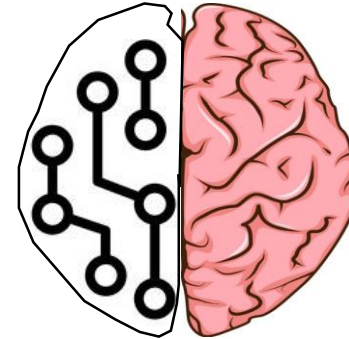


Background

Early Continuous positive airway pressure (CPAP) improves:

- neurological and functional recovery.⁵⁻¹⁰

Therefore there is a need to address sleep apnea early in stroke.



Background



Berlin Questionnaire
SLEEP EVALUATION IN PRIMARY CARE

1. Complete the following: 2. How often do you feel tired or fatigued after waking up?

STOP-Bang questionnaire
Please answer the following questions by checking "yes" or "no" for each one.

	Yes	No
Snoring (Do you snore loudly?)	<input type="checkbox"/>	<input type="checkbox"/>
Tiredness (Do you often feel tired, fatigued, or sleepy during the daytime?)	<input type="checkbox"/>	<input type="checkbox"/>
Observed Apnea (Has anyone observed that you stop breathing, or choke or gasp during your sleep?)	<input type="checkbox"/>	<input type="checkbox"/>
High Blood Pressure (Do you have or are you being treated for high blood pressure?)	<input type="checkbox"/>	<input type="checkbox"/>
BMI (Is your body mass index more than 35 kg per m ² ?)	<input type="checkbox"/>	<input type="checkbox"/>
Age (Are you older than 50 years?)	<input type="checkbox"/>	<input type="checkbox"/>
Neck Circumference (Is your neck circumference greater than 40 cm [15.75 inches]?)	<input type="checkbox"/>	<input type="checkbox"/>
Gender (Are you male?)	<input type="checkbox"/>	<input type="checkbox"/>

Score 1 point for each positive response.
Scoring interpretation: 0 to 2 = low risk, 3 or 4 = intermediate risk, ≥5 = high risk.

Source: University Health Network, Toronto, Ontario, Canada (www.stopbang.ca/osa/screening/php). Used with permission from Sauk Prairie Healthcare.

Final result: 2 or more positive categories indicates a high likelihood of sleep-disordered breathing.

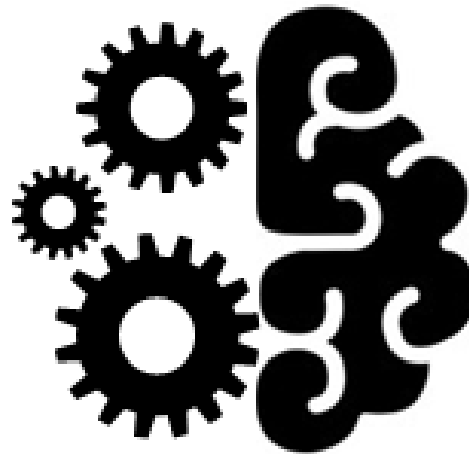


Current methods to screen apnea pose challenges in inpatient rehabilitation

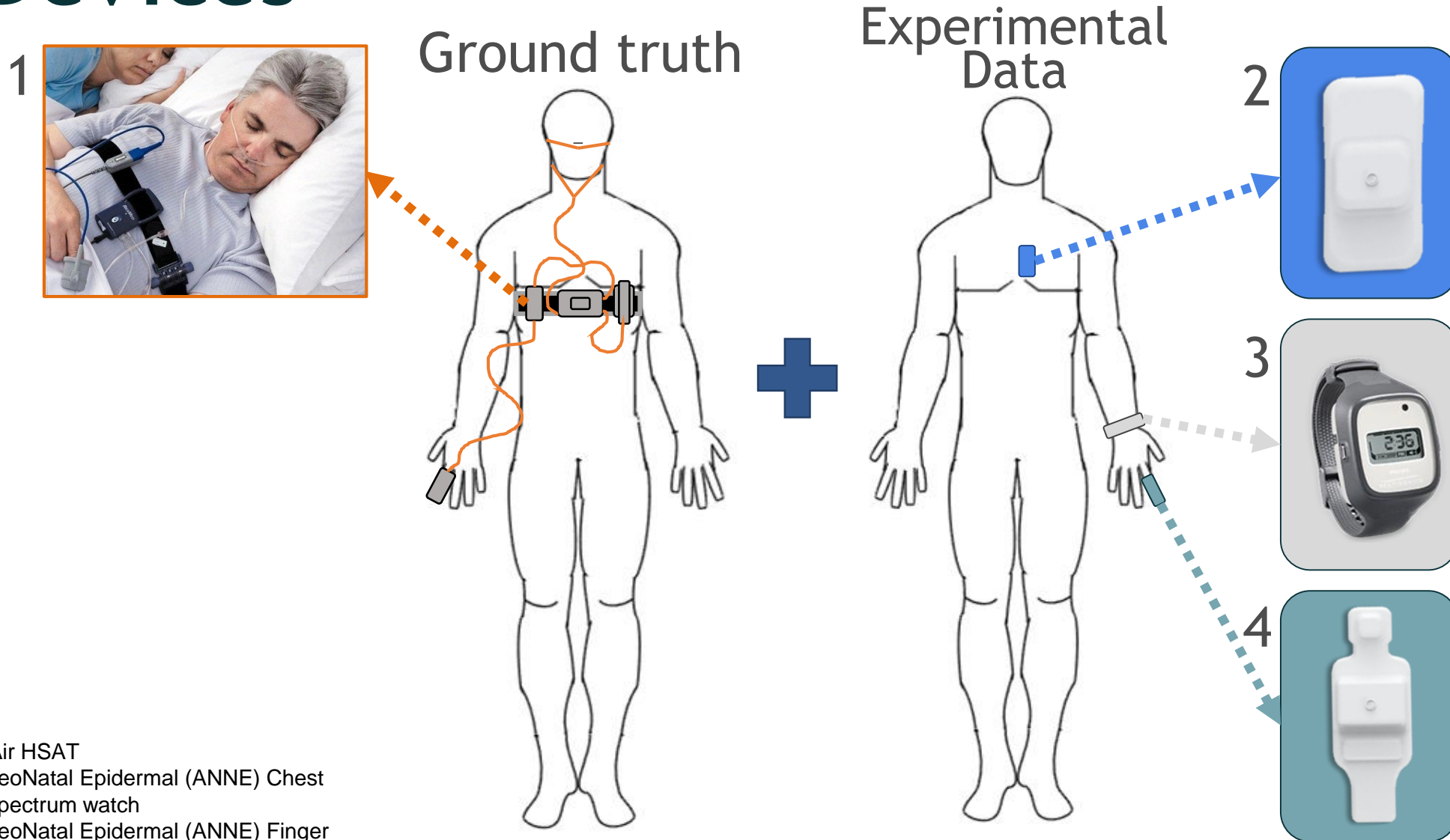
- PSG and HSAT: resource intensive, low patient acceptance/comfort.
- Questionnaires: risk assessment only.

Our Research

This study aimed to **address these limitations** by developing and validating a sleep apnea screening system for individuals with stroke, utilizing **wearable sensors** and **machine learning** which could be worn by patients during early hospitalization and post-acute care



Devices



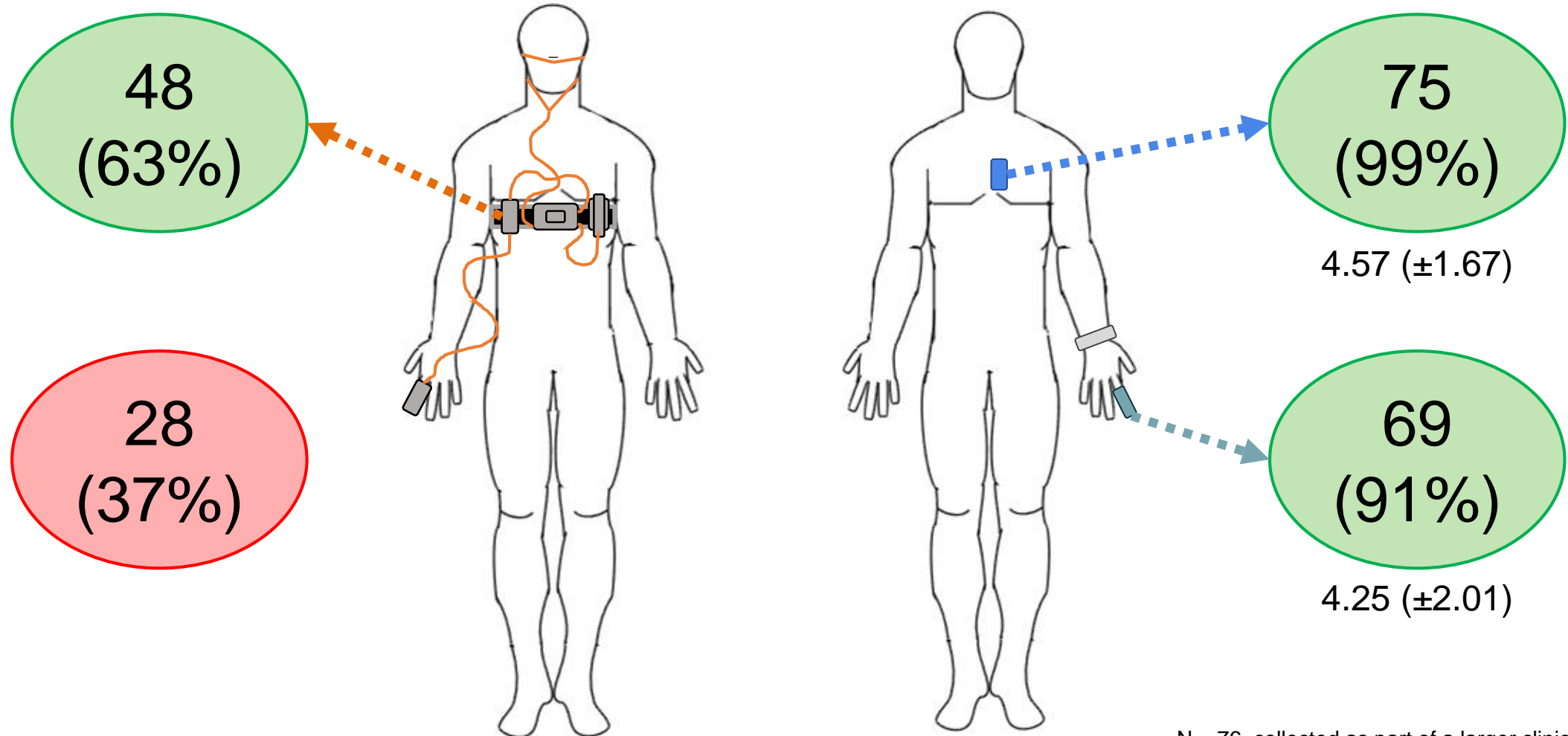
¹ApneaLink Air HSAT

²Advanced NeoNatal Epidermal (ANNE) Chest

³ActiWatch Spectrum watch

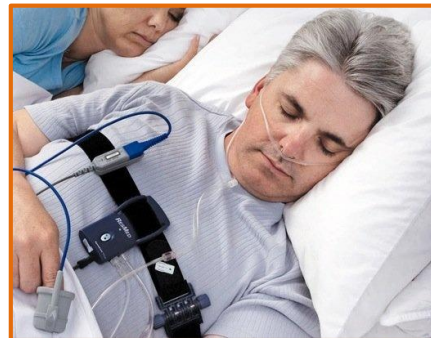
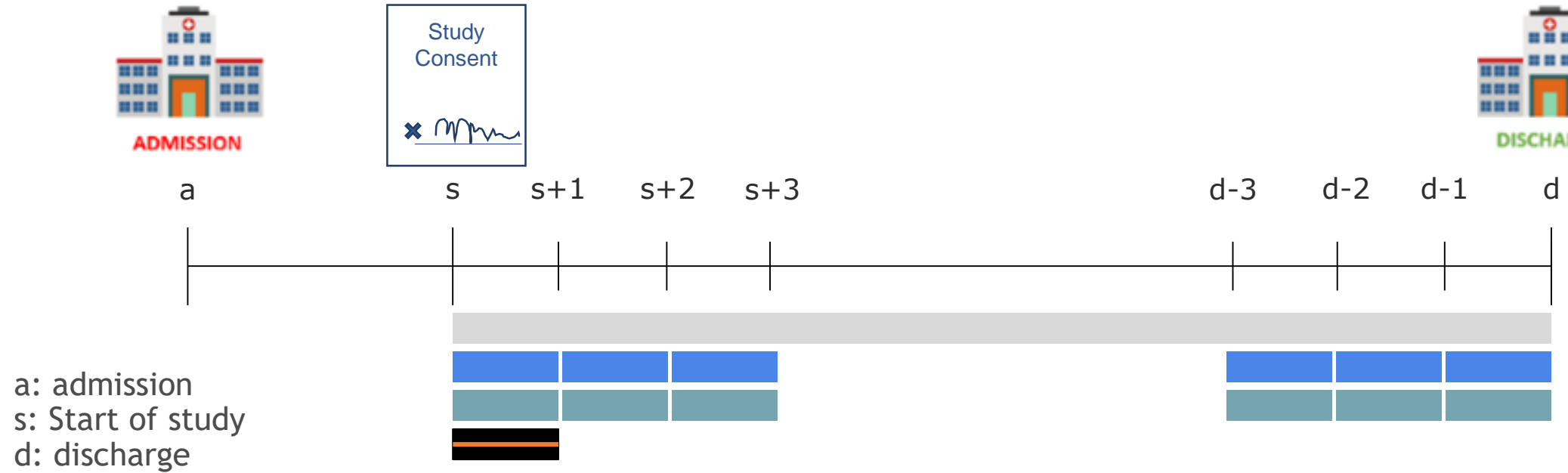
⁴Advanced NeoNatal Epidermal (ANNE) Finger

Device Acceptance (N = 76)



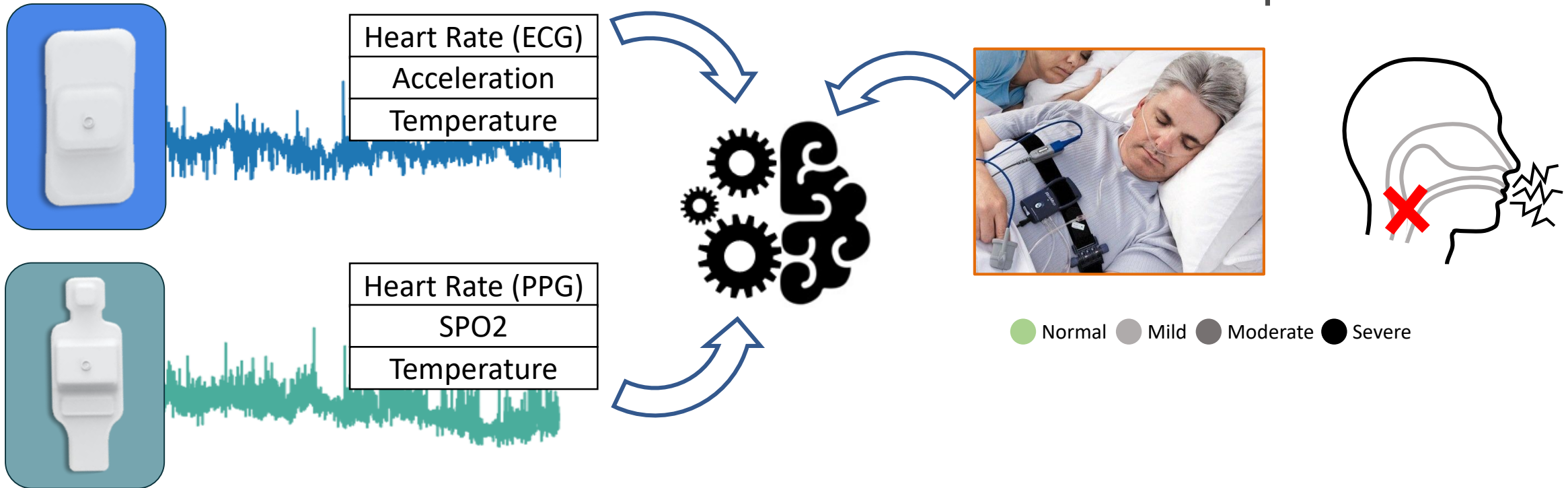
N = 76, collected as part of a larger clinical trial,
SIESTA-Rehab (ClinicalTrials.gov: NCT04254484)

Device Timeline



Devices for Machine Learning

N = 48 with ApneaLink



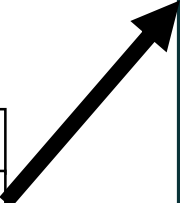
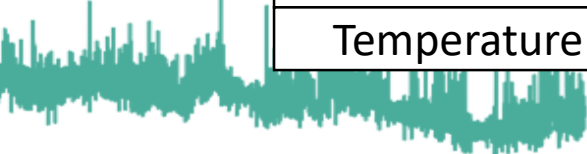
Machine Learning: Feature Design



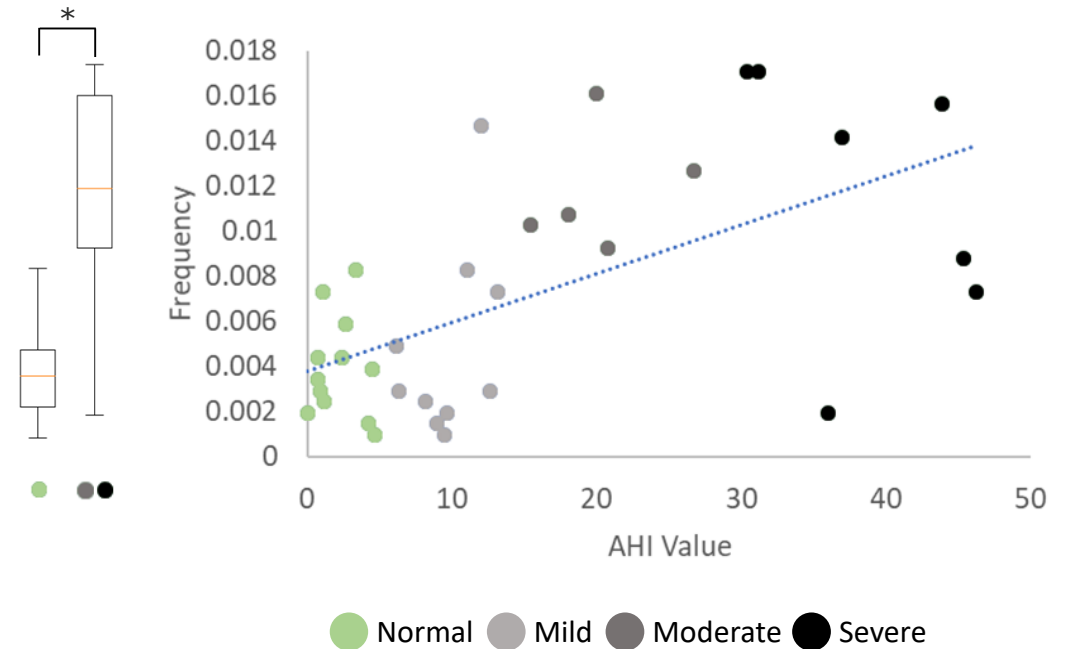
Heart Rate (ECG)
Acceleration
Temperature



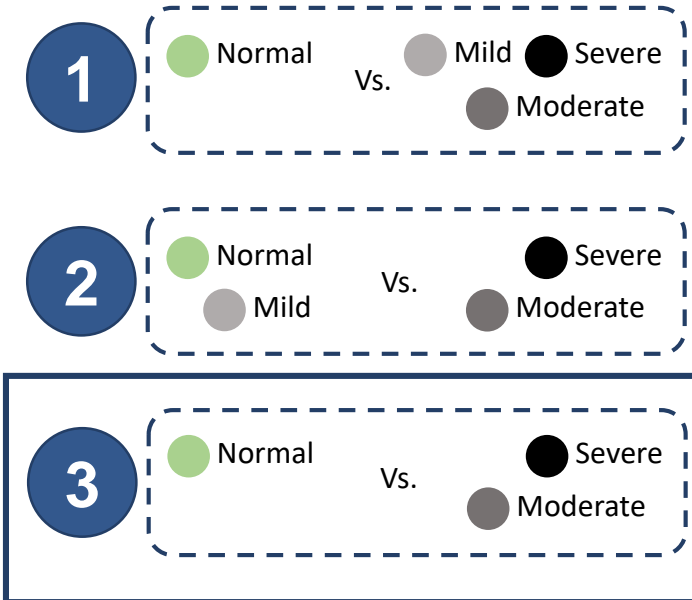
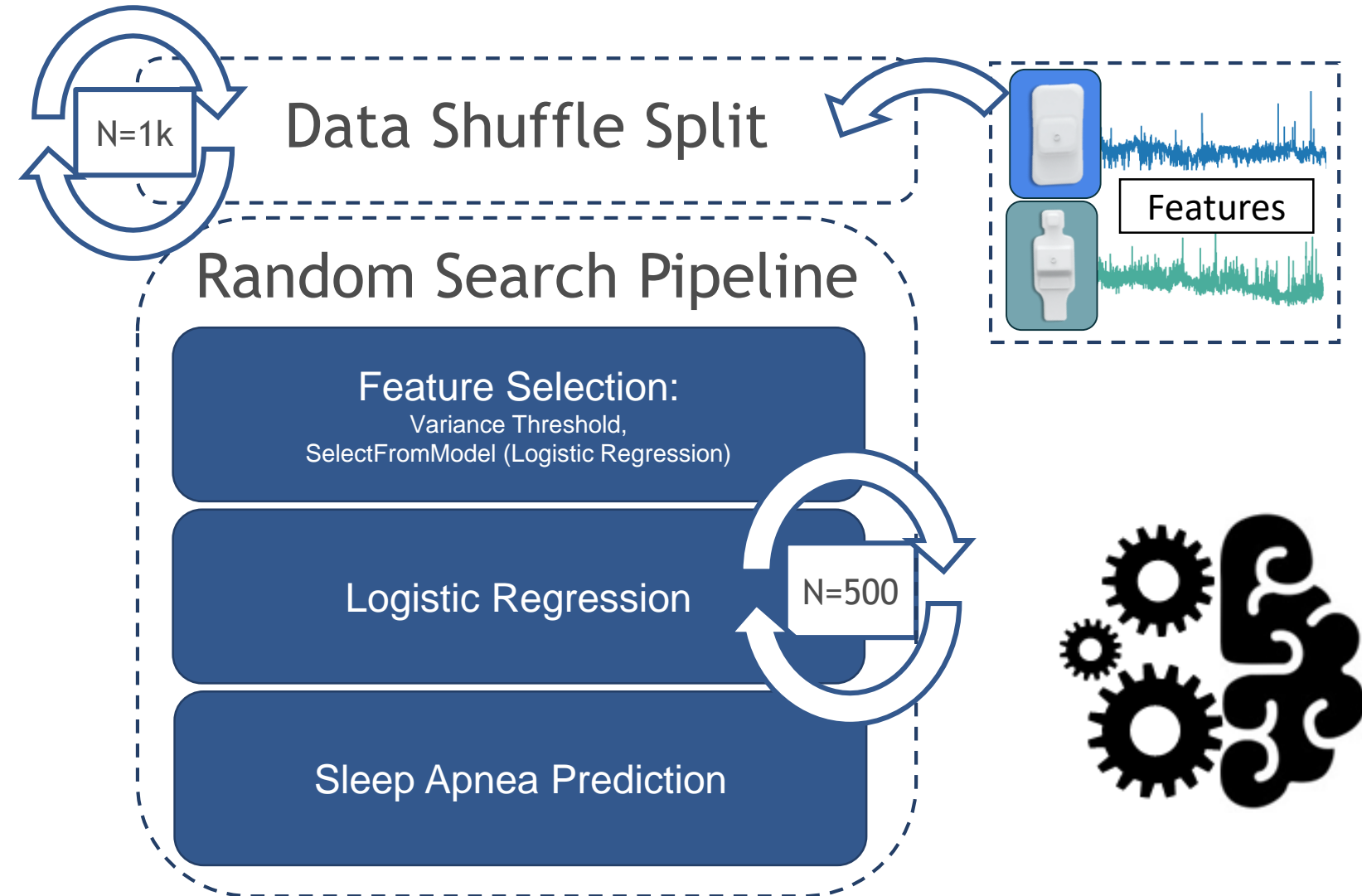
Heart Rate (PPG)
SPO2
Temperature



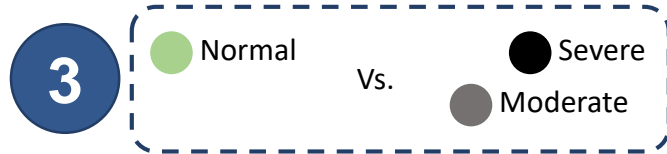
Feature Design:
Example SpO2 PSD median frequency feature



Machine Learning: Pipeline



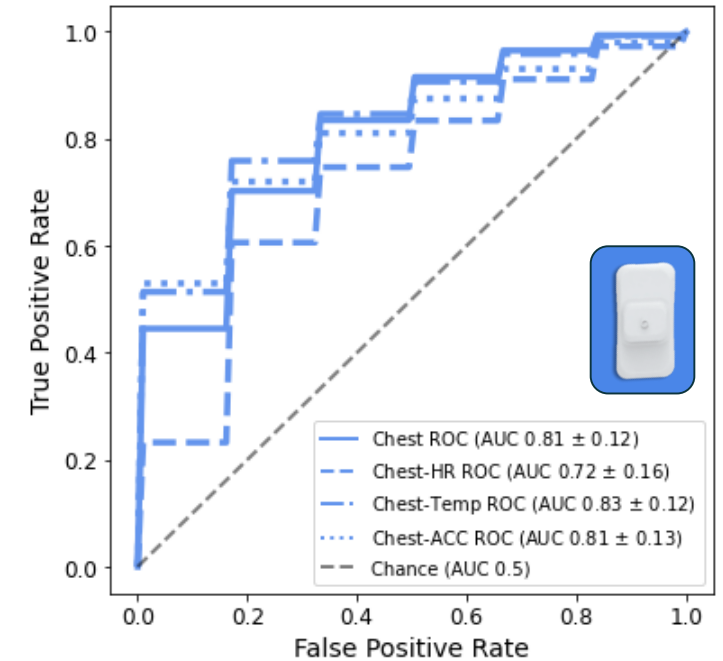
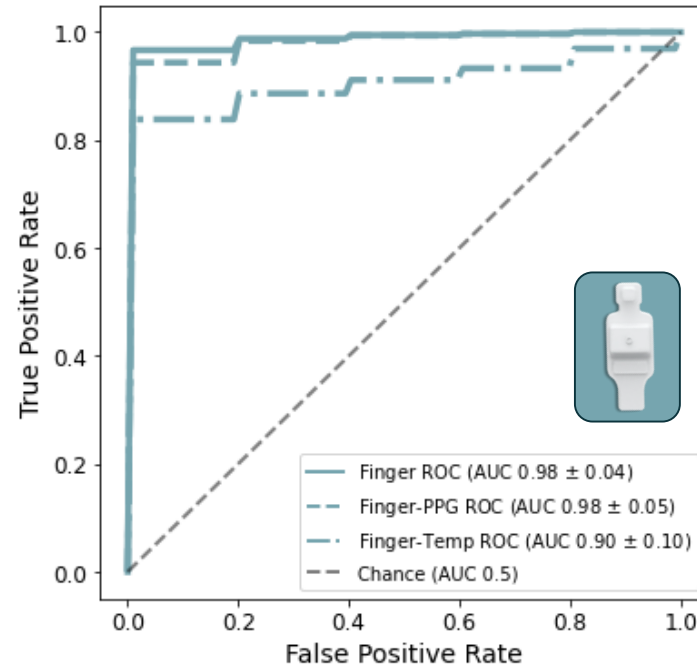
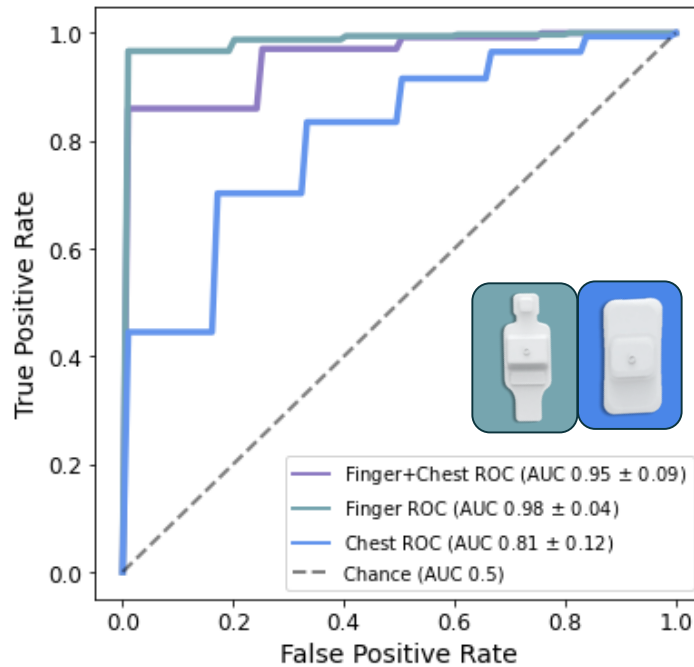
Machine Learning: Results



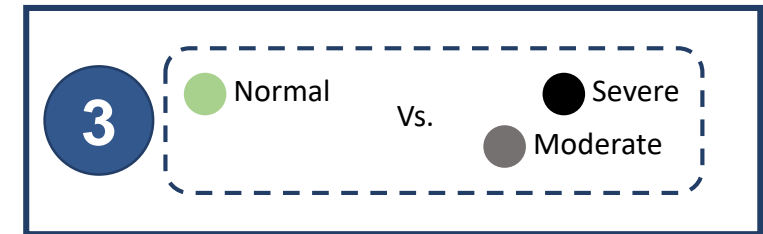
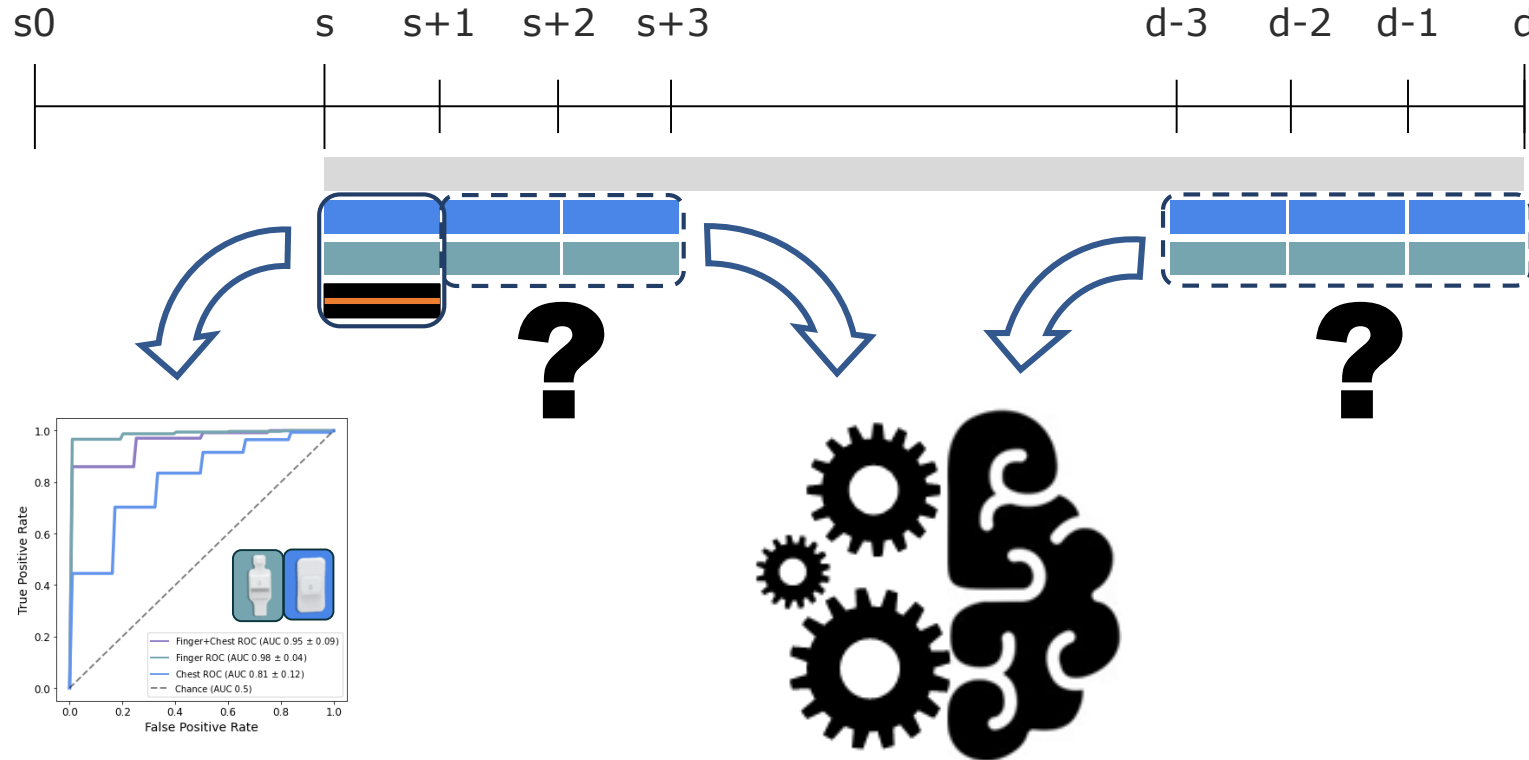
Best model:



Finger derived features

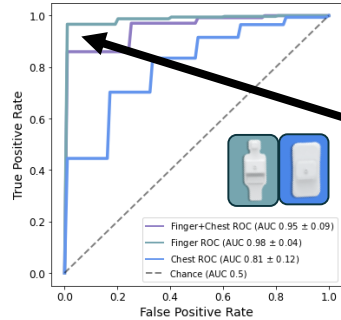
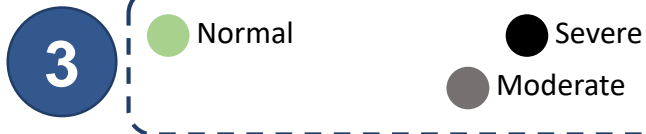


Model Deployment in Acute Care



Model Deployment in Acute Care-Validation

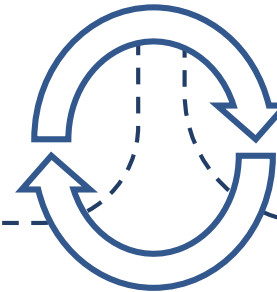
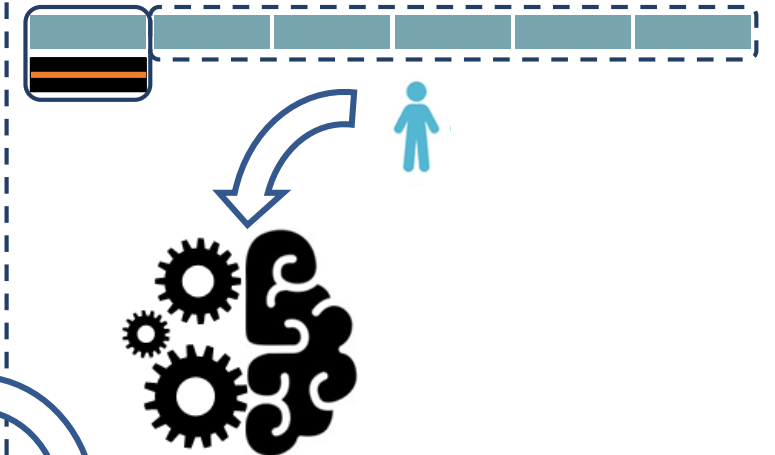
Best Performing Model: *Finger Sensor, set 3*



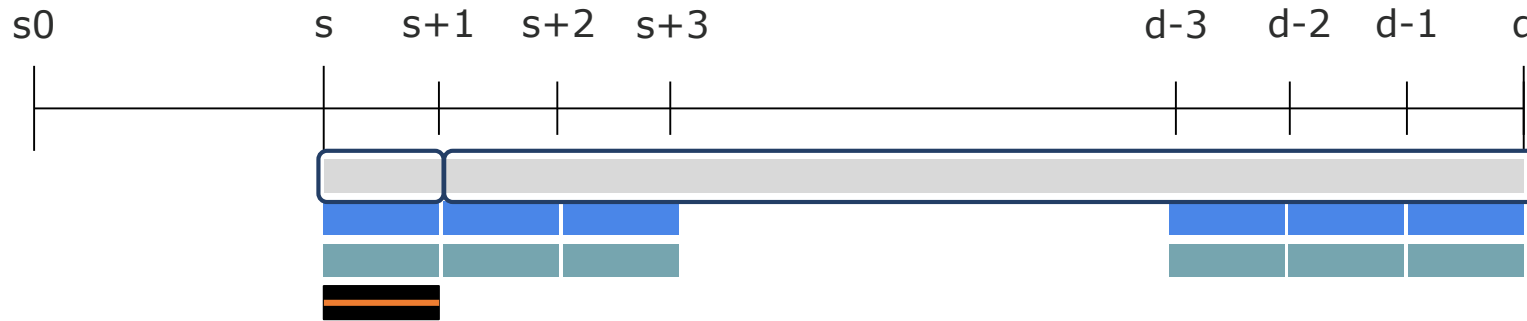
Retrain:
on all but one



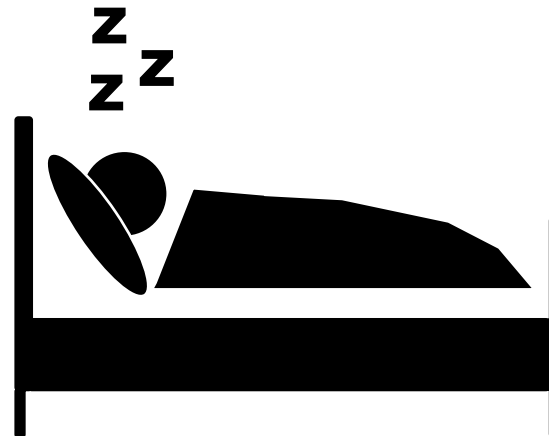
Validate:
on the one excluded



Impact of Abnormal Sleep

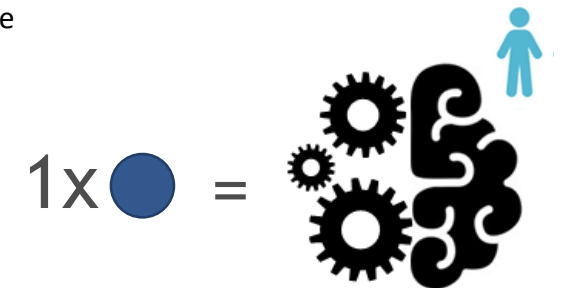
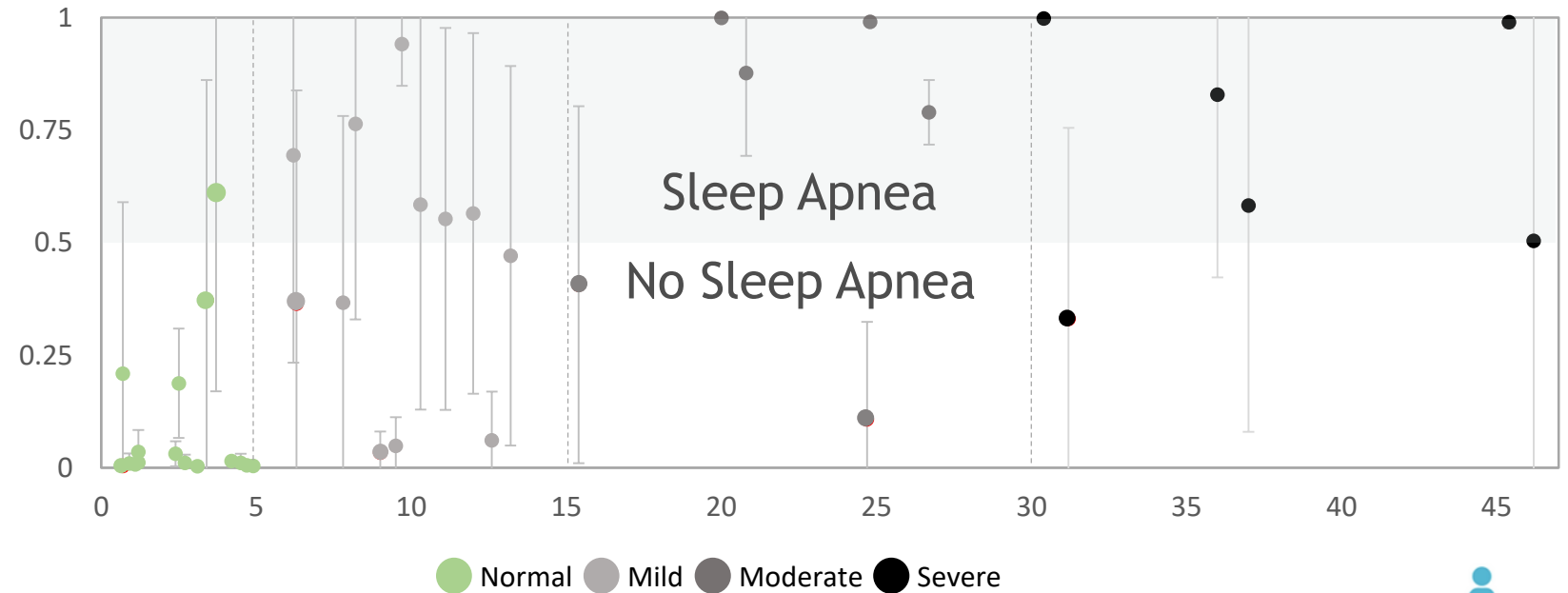


N = 8 with
abnormal
sleep



- ↓ Efficiency
- ↑ Activity Count
- ↑ Wake After Sleep Onset
- ↑ Fragmentation

Model Deployment in Acute Care- Results

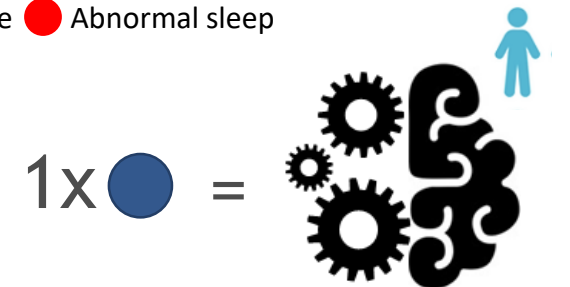
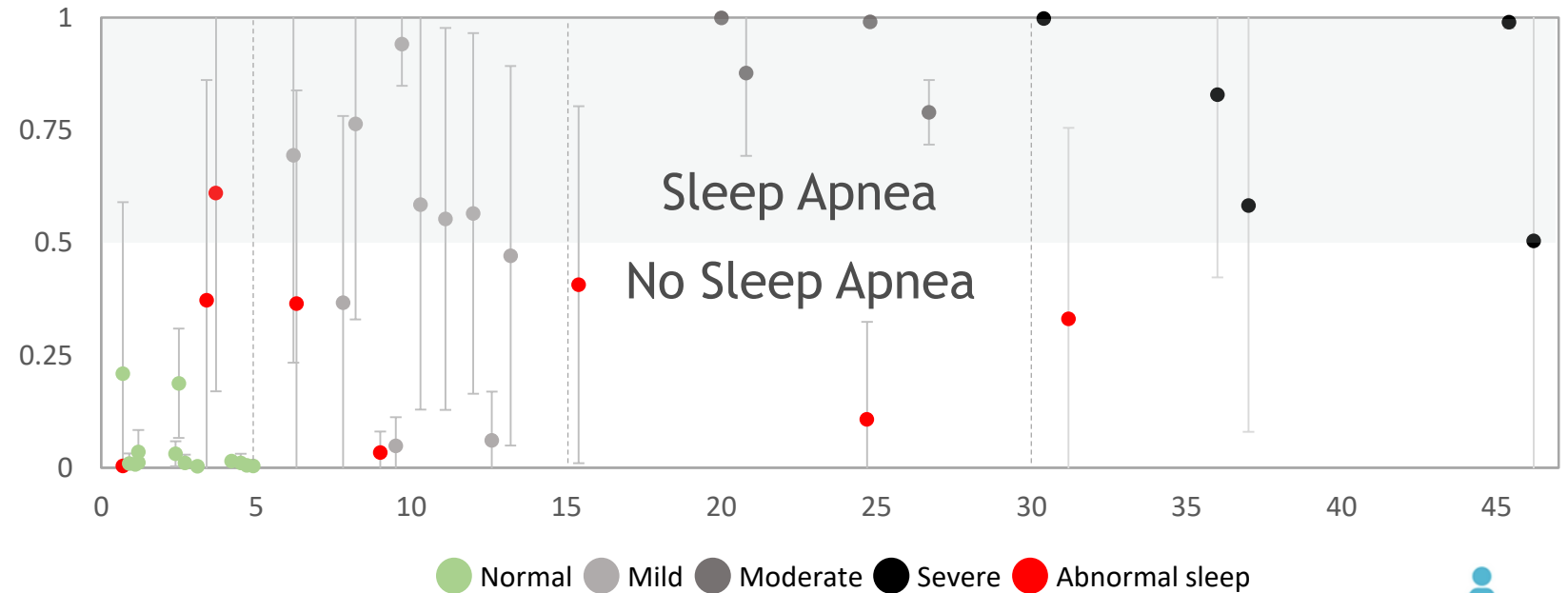


Model Deployment in Acute Care- Results

SLEEP
2024



- ↓ Efficiency
- ↑ Activity Count
- ↑ Wake After Sleep Onset
- ↑ Fragmentation



Conclusion

This work highlights the potential of integrating easy-to-use, non-invasive wearable technology, continuous recordings, and automated analyses into sleep apnea screening, offering a promising avenue for improving sleep, patient care, and outcomes.



References

1. Stickgold R, Walker MP. Sleep-dependent memory triage: evolving generalization through selective processing. *Nat Neurosci*. 2013;16(2):139-145.
2. Nissen C, Kloepper C, Feige B, et al. Sleep-related memory consolidation in primary insomnia. *Journal of sleep research*.. 2011-03;20(1pt2).
3. Dimyan MA, Cohen LG. Neuroplasticity in the context of motor rehabilitation after stroke. *Nat Rev Neurol*. 2011;7(2):76-85.
4. Johnson KG, Johnson DC. Frequency of Sleep Apnea in Stroke and TIA Patients: A Meta-analysis. *Journal of Clinical Sleep Medicine*. 2010;06(02):131-137.
5. Aaronson JA, Hofman WF, Bennekoum CA, et al. Effects of Continuous Positive Airway Pressure on Cognitive and Functional Outcome of Stroke Patients with Obstructive Sleep Apnea: A Randomized Controlled Trial. *Journal of Clinical Sleep Medicine*. 2016;12(04):533-541.
6. Khot SP, Davis AP, Crane DA, et al. Effect of Continuous Positive Airway Pressure on Stroke Rehabilitation: A Pilot Randomized Sham-Controlled Trial. *Journal of Clinical Sleep Medicine*. 2016;12(07):1019-1026.
7. Gupta A, Shukla G, Afsar M, et al. Role of Positive Airway Pressure Therapy for Obstructive Sleep Apnea in Patients With Stroke: A Randomized Controlled Trial. *Journal of Clinical Sleep Medicine*. 2018;14(04):511-521.
8. Javaheri S, Peker Y, Yaggi HK, Bassetti CL. Obstructive sleep apnea and stroke: The mechanisms, the randomized trials, and the road ahead. *Sleep Medicine Reviews*. 2022;61:101568.
9. Minnerup J, Ritter MA, Wersching H, et al. Continuous positive airway pressure ventilation for acute ischemic stroke: a randomized feasibility study. *Stroke*. 2012;43(4):1137-9.
10. Bravata DM, Concato J, Fried T, et al. Continuous positive airway pressure: evaluation of a novel therapy for patients with acute ischemic stroke. *Sleep*. 2011;34(9):1271-7.