

Comparison of Machine Learning Methods to Predict Sleep Quality from Daytime Activity and Nightly Bedroom Environmental Conditions

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Objective: Assess relationship between IAQ/activity with sleep by predicting 7 sleep metrics.

Methods

Smartphone



GPS | Self-report Sleep

BEVO Beacon

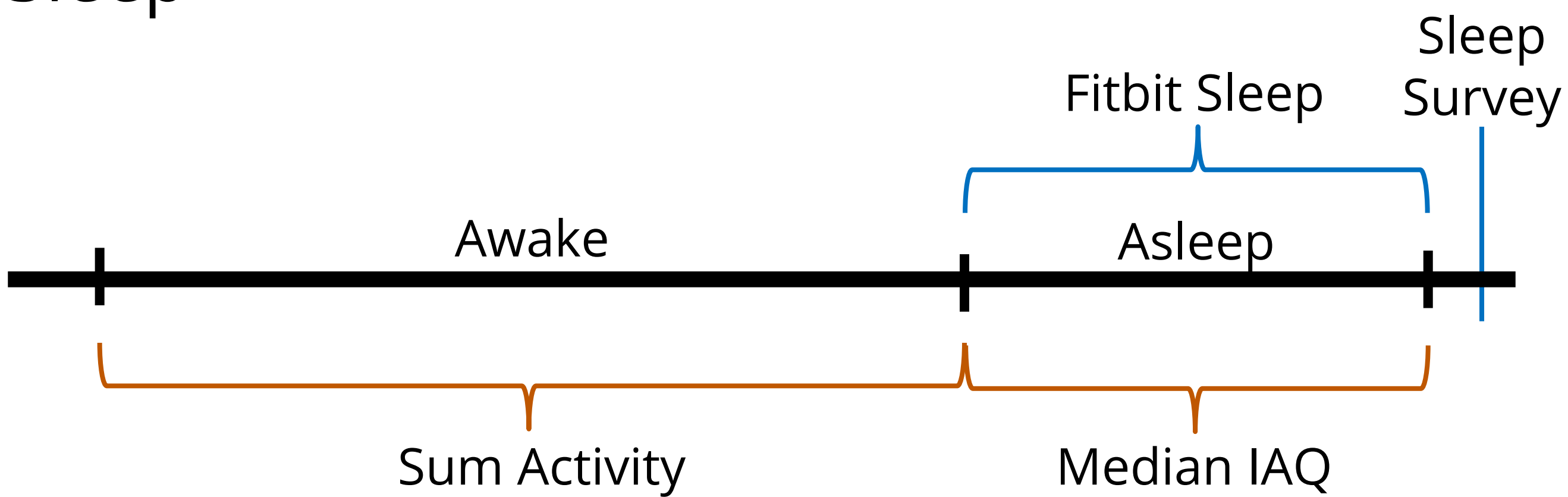


IAQ

Fitbit Inspire HR



Activity | Sleep



Features

PM_{2.5} > 1000 Steps
CO₂ > 30 Minutes
TVOCs Activity
CO
T
RH

Targets

Fitbit {
Sleep Efficiency (SE)
% Rapid-Eye-Movement (REM)
Total Sleep Time (TST)
Survey {
TST
Sleep Onset Latency (SOL)
Number of Awakenings (NAW)
Restfulness

Models

Logistic Regression
K-Nearest Neighbor
k=7, Euclidean
Random Forest
trees=10, depth=1

Results

	Model	Accuracy		F1 (threshold)	Feature Importance
		Train	Test		
SE	LR	0.79	0.78	0.79	0.89 (0.45)
	kNN	0.82	0.74		0.89 (0.26)
	RF	0.79	0.79		0.89 (0.71)
% REM	LR	0.64	0.63	0.64	0.54 (0.28)
	kNN	0.69	0.60		0.53 (0.09)
	RF	0.64	0.64		0.53 (0.34)
TST (Fitbit)	LR	0.65	0.64	0.65	0.52 (0.25)
	kNN	0.70	0.59		0.52 (0.17)
	RF	0.65	0.65		0.52 (0.32)
SOL	LR	0.60	0.51	0.56	0.71 (0.33)
	kNN	0.67	0.54		0.72 (0.23)
	RF	0.58	0.53		0.71 (0.5)
NAW	LR	0.59	0.43	0.52	0.67 (0.36)
	kNN	0.69	0.47		0.69 (0.23)
	RF	0.57	0.40		0.68 (0.48)
Restful	LR	0.83	0.82	0.82	0.9 (0.53)
	kNN	0.83	0.79		0.9 (0.46)
	RF	0.82	0.82		0.9 (0.75)
TST (Survey)	LR	0.69	0.61	0.64	0.78 (0.42)
	kNN	0.70	0.61		0.78 (0.26)
	RF	0.64	0.64		0.78 (0.56)

LR (coefficient)		RF (impurity)	
RH		Steps	
T		TVOC	
Active		Active	
PM _{2.5}		PM _{2.5}	
Active		Active	
TVOC		RH	
T		Active	
Steps		CO ₂	
CO		CO ₂	
T		CO	
CO		CO	
RH		RH	
Active		Active	
RH		RH	

Conclusion: IAQ/activity are related to and can predict sleep quality better than a dummy model that assumes the majority outcome.