Digital Biomarker Applications in Public Health

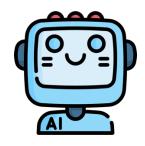
Shravan Aras

The University of Arizona

Center for Biomedical Informatics and Biostatistics

2025-6-12

Let's Get Started



Al Assisted



Code Available



Study Data

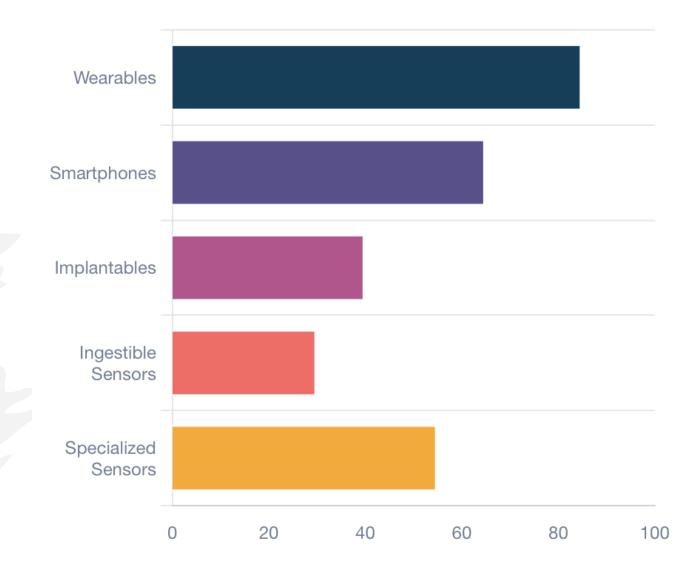


Digital Biomarkers?

One Such Definition

- 1. Digital biomarkers are quantifiable data points.
- 2. Collected by digital devices like <u>wearables</u>, <u>smartphones</u>, <u>or</u> <u>implantable sensors</u>.
- 3. They provide a <u>continuous</u> view of our health.
- 4. They can be collected remotely.

Different Ways to Collect Them

















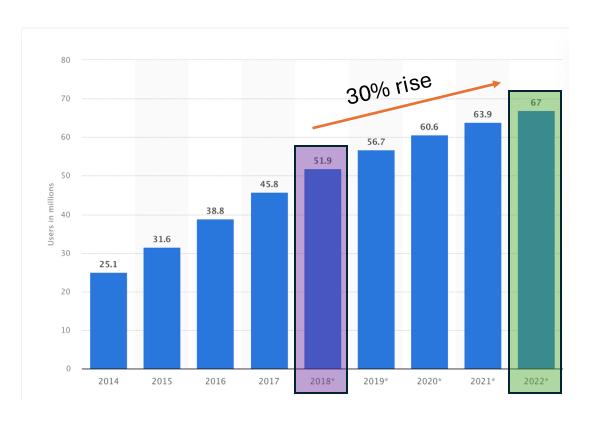


We are going to focus on digital markers coming from wearable devices today!

Why Now?

30% increase in wearable use in last 5 years

Wide range of affordability - \$100 - \$400



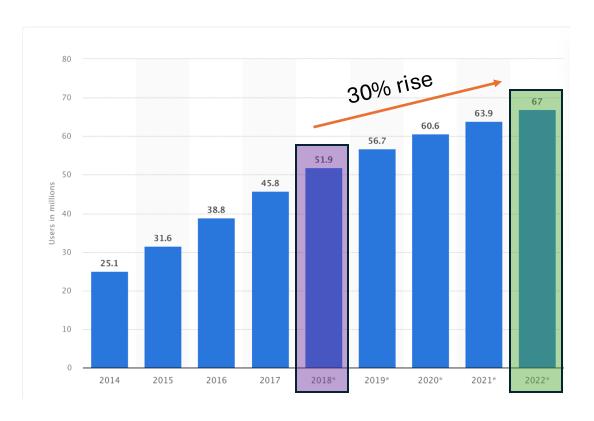
Number of users in millions using wearable fitness devices in U.S.A.

Source: Statista

30% increase in wearable use in last 5 years

Wide range of affordability - \$100 - \$400

More reliable and fashionable



Number of users in millions using wearable fitness devices in U.S.A.

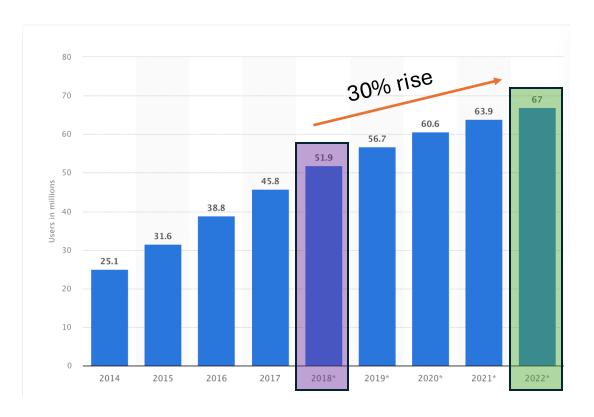
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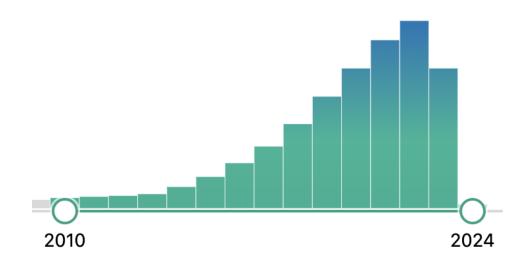
Increased ease of operations



Number of users in millions using wearable fitness devices in U.S.A.

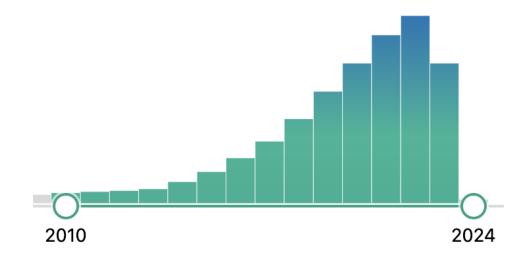
Source: Statista

34% increase in PubMed publications that involve wearable sensors



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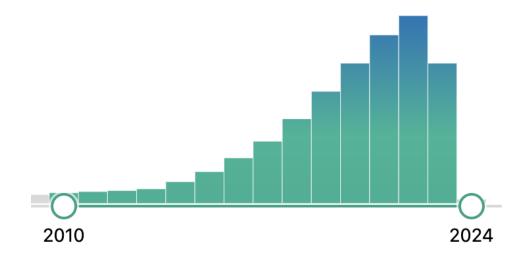
Over \$58M NIH proposals with wearable sensors funded in last 3 years



34% increase in PubMed publications that involve wearable sensors

Over \$58M NIH proposals with wearable sensors funded in last 3 years.

More data points for Idiographic (person specific) and nomothetic (within group)

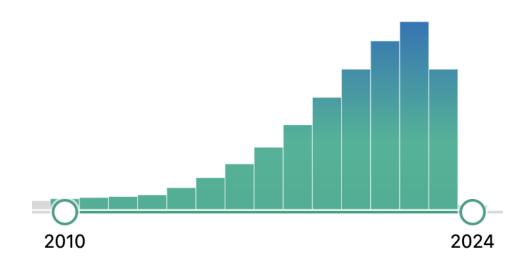


34% increase in PubMed publications that involve wearable sensors

Over \$58M NIH proposals with wearable sensors funded in last 3 years.

More data points for Idiographic (person specific) and nomothetic (within group)

Inferred vs measured (e.g., physical activity, stress, sleep)



Step 1 – Understanding what to measure

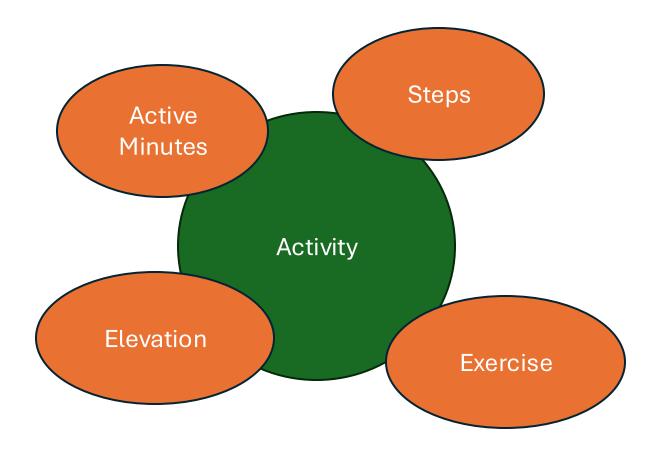


Can feel like a wine menu!

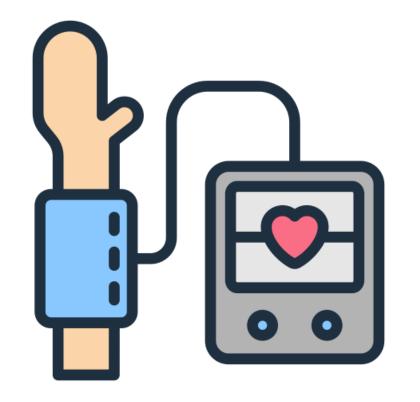
Way too many options out there

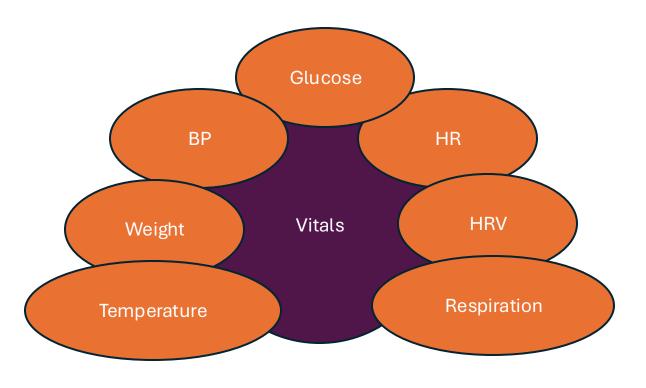
Similar but not the same





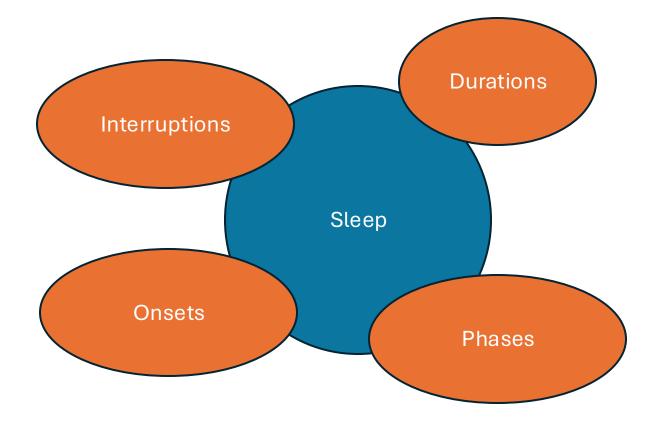
What do we want to measure?





What do we want to measure?



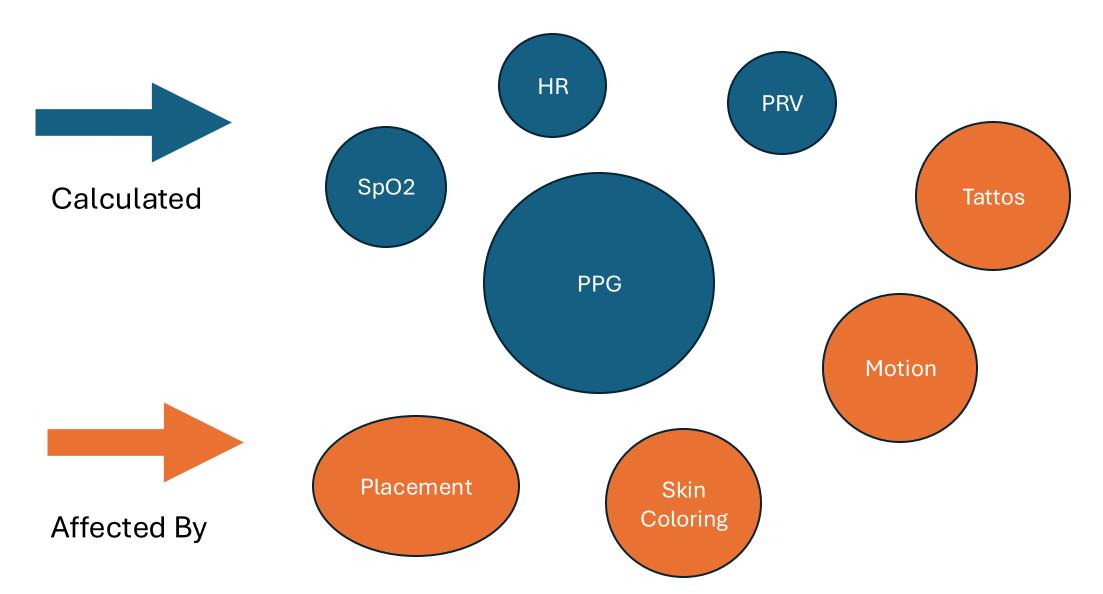


What do we want to measure?

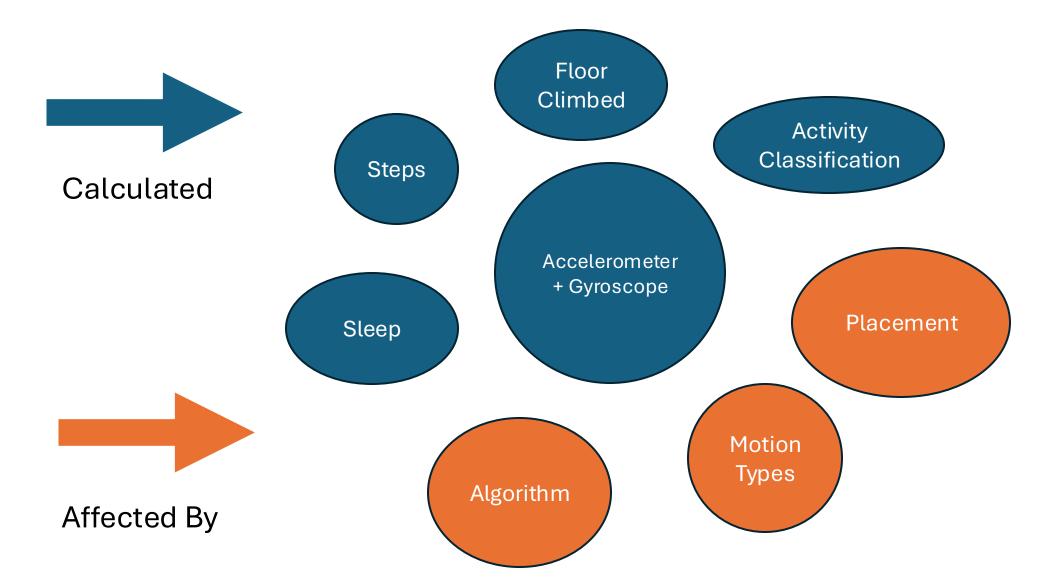
But Wait!!

Not all that you see is calculated, some is estimated!

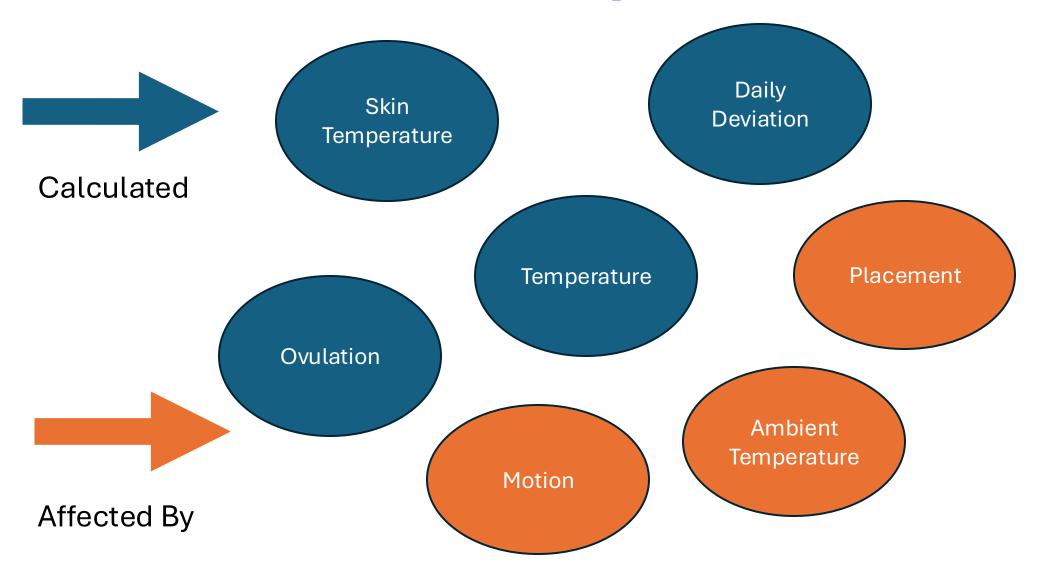
Calculated From PPG Sensor



Calculated From Accelerometer



Calculated From Temperature Sensor



VO2 Max

VO2 Max

• Sleep Phases – REM, Deep, etc.

VO2 Max

Sleep Phases – REM, Deep, etc.

Optical Blood Pressure

VO2 Max

Sleep Phases – REM, Deep, etc.

Optical Blood Pressure

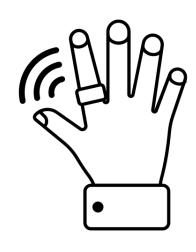
Stress Metric / Body Battery / Recovery Index

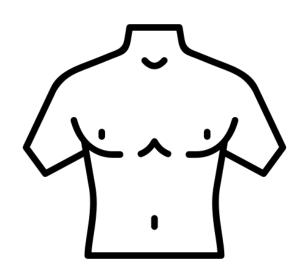
Step 2 – Pick a device

Bring out the roller deck of cards!

Positioning (Common Ones)







Wrist

Finger

Torso

Choosing a wearable

Measurement	Fitbit (\$)	Apple Watch (\$\$)	Oura (\$\$)	Empatica (\$\$\$)
Daily Activity	Standard	Standard	Standard	Raw Acceleration
Exercise Sessions	Standard	Standard	Standard	Basic
Intraday HR	~ 5 seconds	~ 5 minutes	~ 5 minutes	~ 1minute
Resting HR	Yes	Yes	Yes	Can Calculate
Sleep Duration	Yes	Yes	Yes	Yes
Sleep Phases	Yes	Yes	Yes	No
HRV	Day / Night (summary / intra)	Day / Night (summary / intra)	Night (5 mins)	1 min (RMSSD)
IBI	No	No	Yes	Yes
Raw Acceleration	No	Yes	No	Yes
Temperature	Summary	Summary / ~5s (SensorKit)	Summary / 5m(Research)	1 min
spo2	Yes	Yes	Yes	? 30

Activity, Exercise, Sleep

	Measurement	Fitbit (\$)	Apple Watch (\$\$)	Oura (\$\$)	Empatica (\$\$\$)
Γ	Daily Activity	Standard	Standard	Standard	Raw Acceleration
	Exercise Sessions	Standard	Standard	Standard	Basic
	Intraday HR	~5 sec	Everyone supports the basics		~ 1minute
	Resting HR	Yes	e supports the pasics	S	Can Calculate
ı	Sleep Duration	Yes	Yes	Yes	Yes
	Sleep Phases	Yes	Yes	Yes	No
ı	HRV	Day / Night (summary / intra)	Day / Night (summary / intra)	Night (5 mins)	1 min (RMSSD)
	IBI	No	No	Yes	Yes
	Raw Acceleration	No	Yes	No	Yes
	Temperature	Summary	Summary / ~5s (SensorKit)	Summary / 5m(Research)	1 min
	spo2	Yes	Yes	Yes	?

Intraday Heart Rate (IHR)

Measurement	Fitbit (\$)	Apple Watch (\$\$)	Oura (\$\$)	Empatica (\$\$\$)	
Daily Activity	Standard	Standard	Standard	Raw Acceleration	
Exercise Sessions	Standard	Standard	Standard	Basic	
Intraday HR	~ 5 seconds	~ 5 minutes	~ 5 minutes	~ 1minute	
Resting HK	Yes	Yes	Yes	Can Calculate	
Sleep Duration	Yes Intraday	HR support is variable	s	Yes	
Sleep Phases	Yes	100	s	No	
HRV	Day / Night (summary / intra)	Day / Night (summary / intra)	Night (5 mins)	1 min (RMSSD)	
IBI	No	No	Yes	Yes	
Raw Acceleration	No	Yes	No	Yes	
Temperature	Summary	Summary / ~5s (SensorKit)	Summary / 5m(Research)	1 min	
spo2	Yes	Yes	Yes	?	

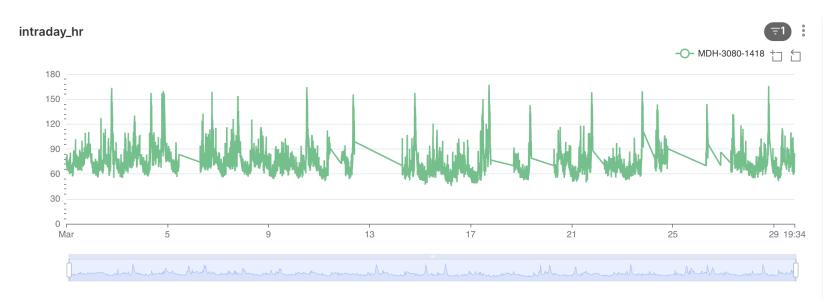
Some devices offer something special

Measurement	Fitbit (\$)	Apple Watch (\$\$)	Oura (\$\$)	Empatica (\$\$\$)
Daily Activity	Standard	Standard	Standard	Raw Acceleration
Exercise Sessions	Standard	Standard	Standard	Basic
Intraday HR	~5 seconds	~ 5 minutes	~ 5 minutes	~ 1minute
Resting HR	Yes	Yes	Yes	Can Calculate
Sleep Duration	Yes	Yes	Yes	Yes
Sleep Phases	Yes			No
HRV	Day /	ecial support	light (5 mins)	1 min (RMSSD)
	(summary / intra)	(summary / intra)		
IBI	No	No	Yes	Yes
Raw Acceleration	No	Yes	No	Yes
Temperature	Summary	Summary / ~5s (SensorKit)	Summary / 5m(Research)	1 min
spo2	Yes	Yes	Yes	?

Fitbit can give more than just summaries

Researchers can request access to Fitbit intraday data to unlock a wide range of high frequency metrics

	type
1	activities-heart
2	activities-distance
3	activities-floors
1	activities-steps
5	activities-elevation
6	br
7	activities-calories
8	spo2
9	hrv



Intraday day HR values for a participant

Get in-depth motion data with SensorKit

Multiple high frequency metrics

- Acceleration
- Rotation Rate
- Light Sensor
- Temperature

Acceleration & Rotation

- About 100Hz sampling rate
- Does affect battery almost half

Does impact battery life

```
{
  "timestamp": 140941.628781,
  "acceleration": {
    "x": -0.30615234375,
    "y": 0.634033203125,
    "z": 0.739990234375
  },
  "startDate": "2023-08-28T06:43:57-0700",
  "identifier": 0
}
```

What about BP, Glucose, Weight, etc.?

- Most BLE / BT smart devices connect Google Fit & Apple Health Kit
- Connecting them can be finicky -> right permissions
- Syncing can be a challenge
- Good study material, instructions and monitoring can help!

	type	startdate	date	value	units	sourceidentifier	sourcename	sour
	BloodGlucose	2023-03-12 18:28:25.000	2023-03-12 18:28:25.000	132	mg/dL	com.dexcom.G6	Dexcom G6	17375
2	BloodGlucose	2023-03-12 18:33:25.000	2023-03-12 18:33:25.000	130	mg/dL	com.dexcom.G6	Dexcom G6	17375
}	BloodGlucose	2023-03-12 18:38:25.000	2023-03-12 18:38:25.000	125	mg/dL	com.dexcom.G6	Dexcom G6	17375
ŀ	BloodGlucose	2023-03-12 18:43:25.000	2023-03-12 18:43:25.000	112	mg/dL	com.dexcom.G6	Dexcom G6	17375
j	BloodGlucose	2023-03-12 18:48:25.000	2023-03-12 18:48:25.000	105	mg/dL	com.dexcom.G6	Dexcom G6	17375
ì	BloodGlucose	2023-03-12 18:53:25.000	2023-03-12 18:53:25.000	98	mg/dL	com.dexcom.G6	Dexcom G6	17375

Gemini Deep Research – Learn about devices

A Universe of Health Insights

Fitbit provides a holistic view of well-being, moving far beyond simple fitness tracking. The API grants access to a rich dataset spanning multiple categories, empowering applications in personal wellness, academic research, and preventative health. This structured access allows for the creation of sophisticated tools that can correlate lifestyle factors with health outcomes.

6

Core Data Categories

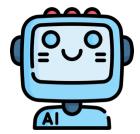
Activity, Sleep, Heart Rate, Body Composition, Nutrition, and Advanced Vitals.

The API is a significant platform for a diverse range of applications, with a strong emphasis on research.

Health & Wellness Apps Medical Research

Personalized Insights

Primary API Use Cases



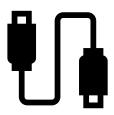
Step 3 – Connecting to Devices

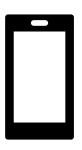
Knock Knock, can I get some data from you?

To Code or Not To









API (JSON) Data Dump (csv, zip, parquet, JSON)

Plug it in

3rd Party Providers

You can use Al to generate most of the code SO

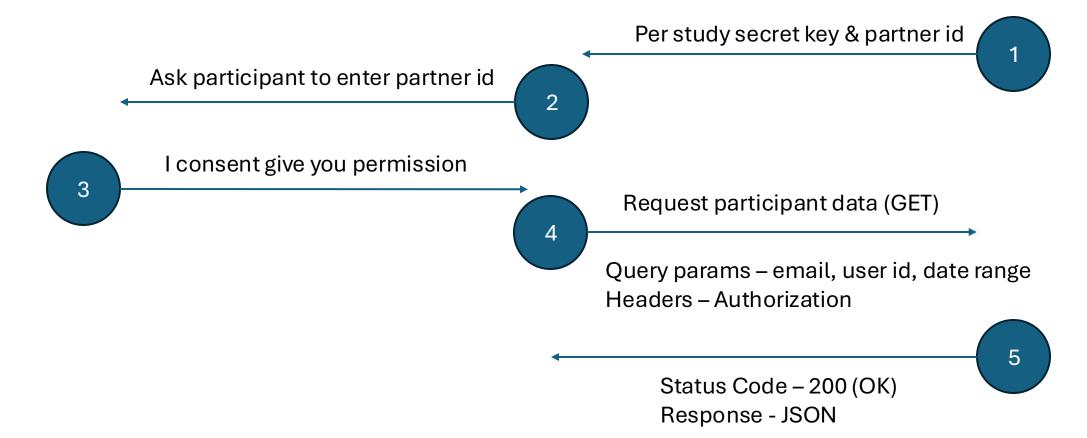
Let's understand some of the fundamental patterns that you can ask AI to code up

Types of API access – secret key, partner id

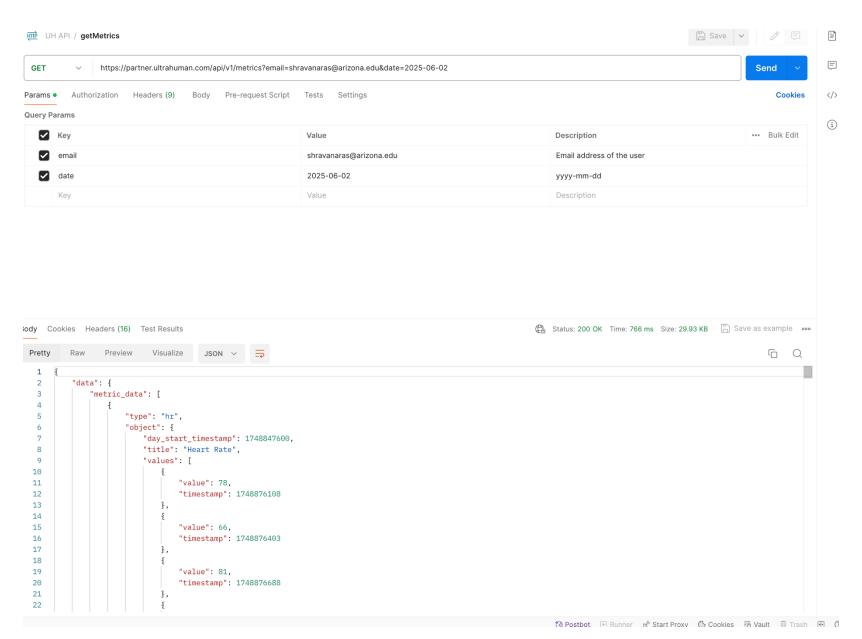
User

Research Team

Device



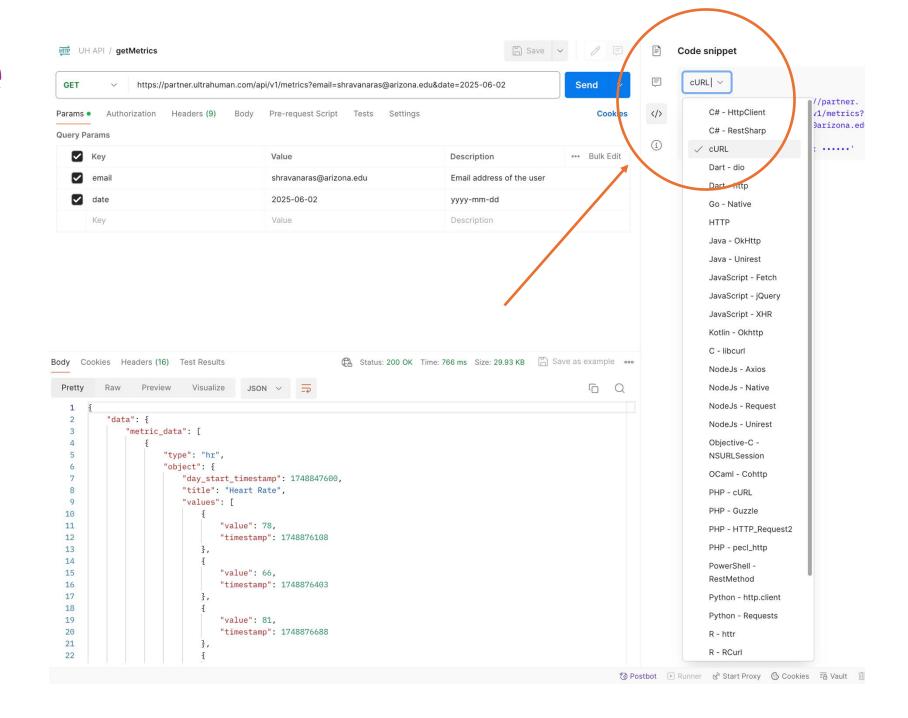
Postman







Generate Code



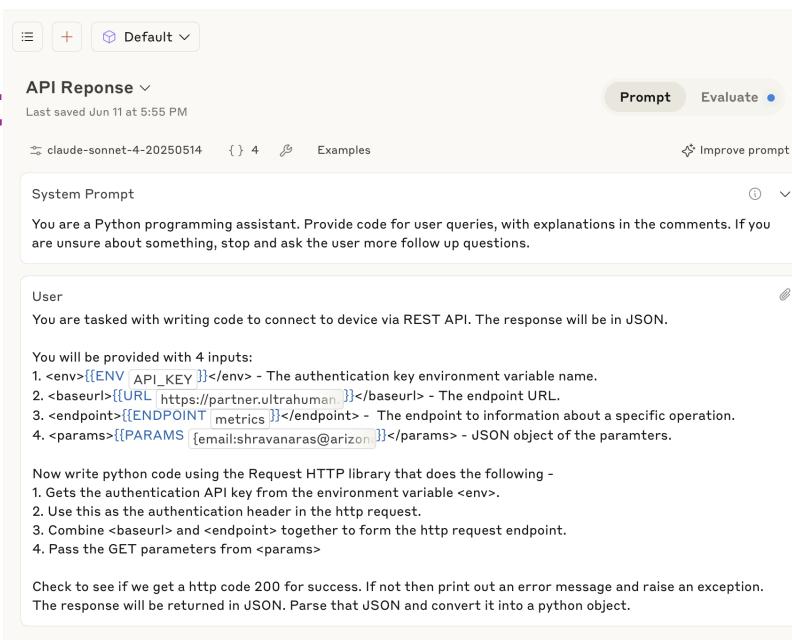


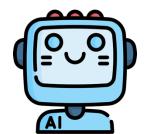


Using your favorite LLM – System Prompt

You are a Python programming assistant. Provide code for user queries, with explanations in the comments. If you are unsure about something, stop and ask the user more follow up questions.

User Prompt









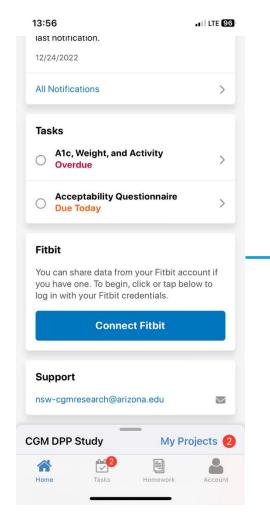
Types of API Access – OAuth 2.0

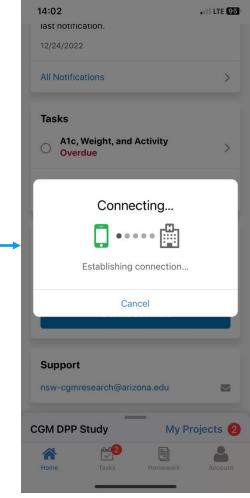
Research Team User Device Create a developer account Get account key (secret) 2 I want to connect my device Give me link the user can go to 2 Developer secret, callback URL Go to this link and login 5 I am logging in and giving the research team access to the following ... 6 Here is a temp token, refresh token Give me participant data (temp token)

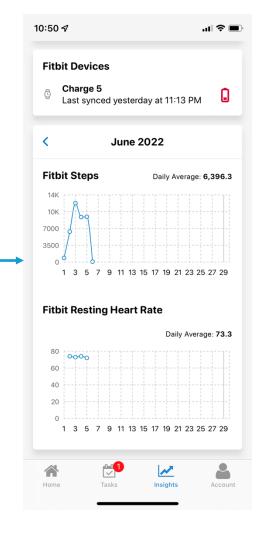
Using 3rd party mHealth Eco-system

User gives

permission







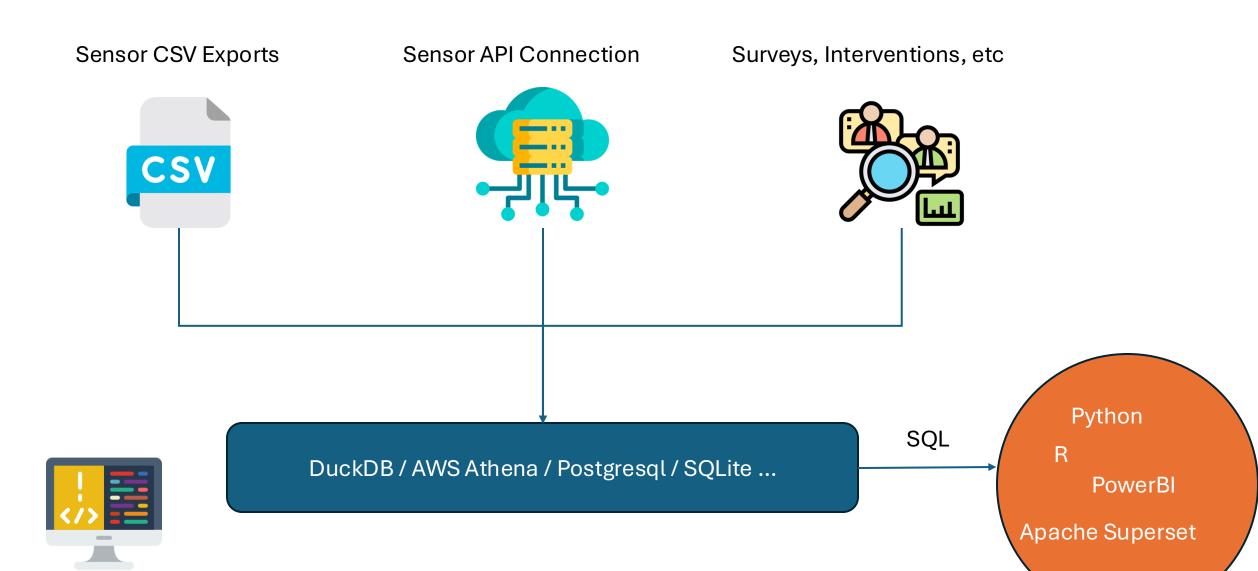
Connected!



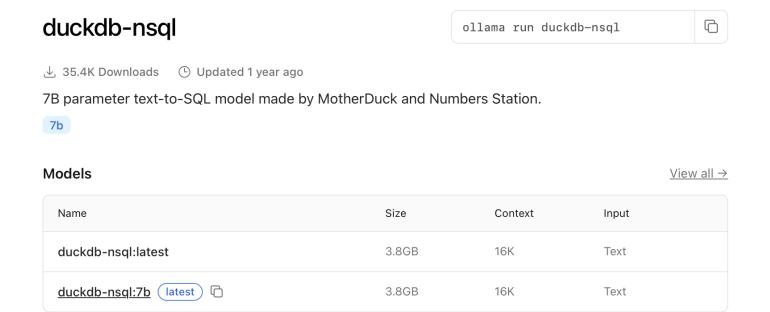
Step 4 – Store data and connect

Time to make sure things are easy to query and connect with other rest of the research study instruments

Using a database solution



Using AI with SQL Generation





Things to watch out for CSV

- Look at date and time very carefully.
 - Keep things as UNIX timestamp (BIGINT).
 - Convert it all to local UTC ISO timestamp 2024-01-19T16:00:00-08:00
- Make sure the CSV files have a header.

- Make sure to reduce duplicates in data ingestion.
 - Creating primary keys.
 - Time based logic.

Brings it all together

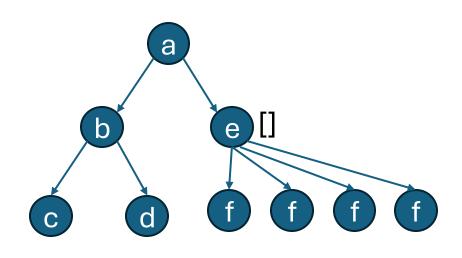
Merge data from different endpoints into a single RDS layer

• Use your favorite programming language – R, Python, C++, etc.

Almost all no code visualization tools support SQ: connectors

Use LLM models to generate SQL queries

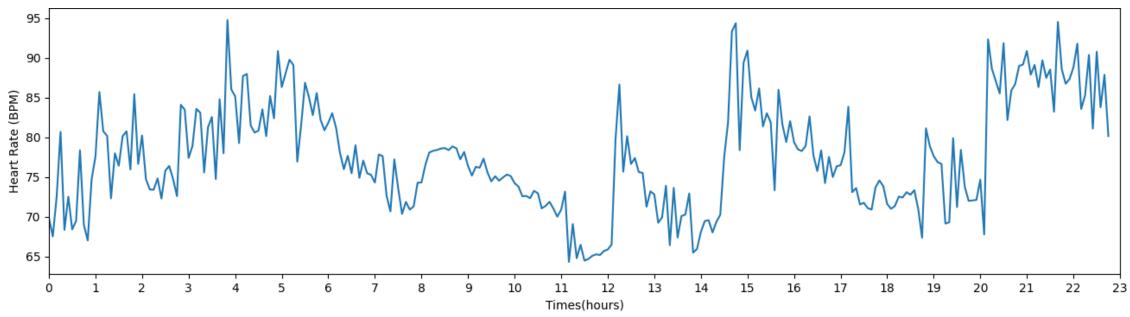
JSON needs a bit more work



- Columns in tabular structure
 - A
 - A_b_c
 - A_b_d
 - Might need filling
 - A_e_f
 - Each element is a row

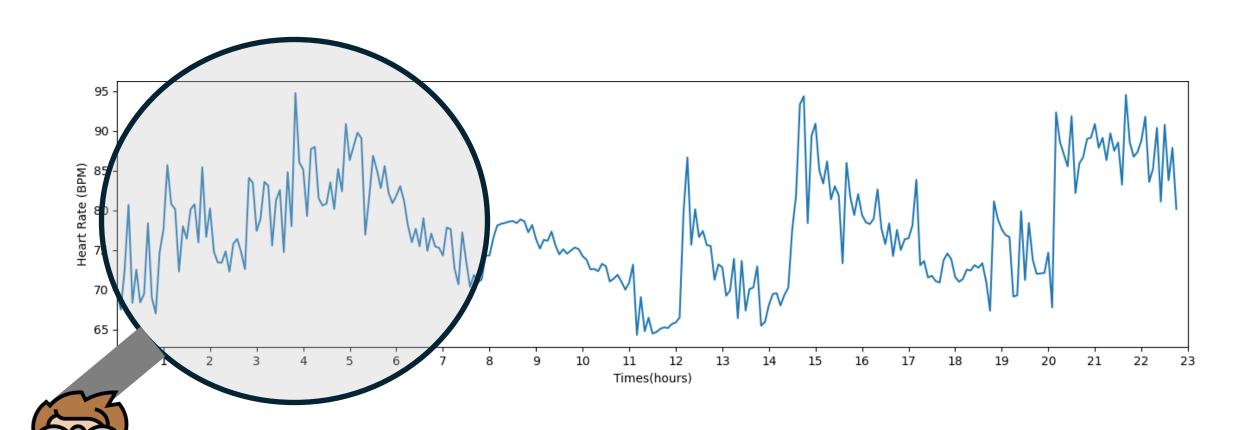


Something is Not Right Here!





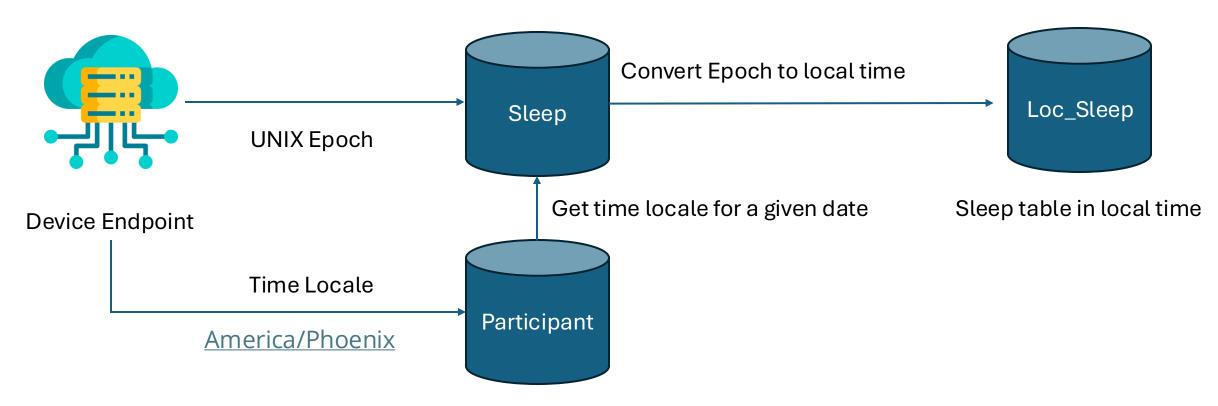
That HR looks odd for sleeping?



We Forgot Time Zones!

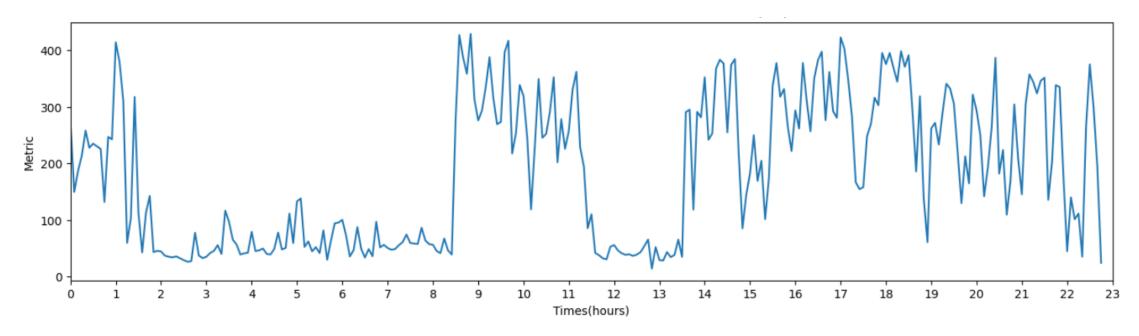
We Forgot Time Zones! Let's adjust for it now!

Time Zones need a special mention



Pid, Date, Time Locale

Much Better!





Example of Our DB

# 🗸	timestamp_ns	datetime_utc	datetime	ibi_ms	hr 5	device_id	▼	timezone ▽	algoversion ▽	
1	1706131308920196992	2024-01-24 21:21:48.920	2024-01-24 13:21:48.920	714.853958	83.933227659348		P3	-28800	{major': 8, 'minor': 12, 'patch': 3}	
2	1706131309631769194	2024-01-24 21:21:49.631	2024-01-24 13:21:49.631	711.572202	84.32032593650982		P3	-28800	{major': 8, 'minor': 12, 'patch': 3}	
3	1706131310294824256	2024-01-24 21:21:50.294	2024-01-24 13:21:50.294	663.055062	90.49022236406665		P3	-28800	{major': 8, 'minor': 12, 'patch': 3}	
4	1706131311000879662	2024-01-24 21:21:51.000	2024-01-24 13:21:51.000	706.055406	84.97916663497652		P3	-28800	{major': 8, 'minor': 12, 'patch': 3}	

Values returned by Empatica

Per day adjusted local time

Extracted local time offset for reference



Step 4.b – Stay on top of participant compliance

Who is naughty and nice!

Inferring Wear Time

Looking at HR values

• Checking to see if step counts stay as default

Looking at resting HR

BE Creative!

Creating a compliance dashboard

Fitbit Compliance App

Overview	Sleep	•	Activity	Resting I	HR H	Heart Rate Variability	
Participant 🕏	Last Synced 🕏	Days Since Sync	\$ Sleep	Activity	Resting HR	≑ B	
filter data							
→ MDH	2023-07-13	68	Yes	Yes	Yes	Y	
MDH	2023-05-14	128	Yes	Yes	Yes	Y	
MDH	2023-03-05	198	Yes	Yes	Yes	ž	
→ MDH	2023-03-20	183	Yes	Yes	Yes	7	
→ MDH	2023-06-03	108	Yes	Yes	Yes		
→ MDH	2023-05-15	127	Yes	Yes	Yes		
→ MDH	2023-06-03	108	Yes	Yes	Yes		
MDH	2023-07-13	68	Yes	Yes	Yes		
MDH	2023-07-13	68	Yes	Yes	Yes		
→ MDH	2023-06-15	96	Yes	Yes	Yes		
MDH	2023-07-03	78	Yes	Yes	Yes		
MDH	2023-08-18	32	Yes	Yes	Yes		
MDH	2023-06-17	94	Yes	Yes	Yes		
MDH	2023-07-13	68	Yes	Yes	Yes		
MDH	2023-07-09	72	Yes	Yes	Yes		
MDH	2023-05-11	131	Yes	Yes	Yes		
MDH	2023-02-13	218	No	Yes	Yes		
→ MDH	2023-03-29	174	Yes	Yes	Yes		

Number of participants who have -

- Fitbit connected 18
- Sleep data 17
- Activity data 18
- Resting HR data 18
- HRV data 16
- Not synced in the last 2 days or more 18

Legends

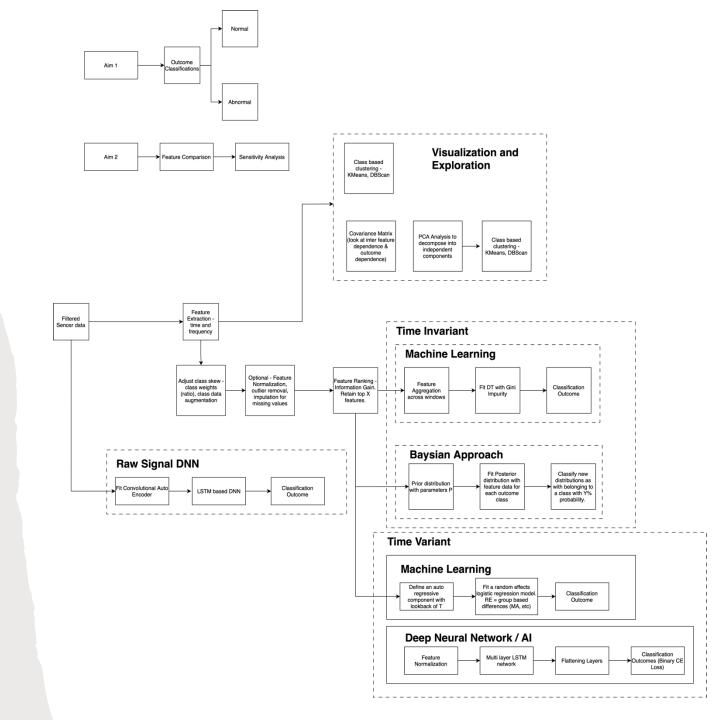
• Green - Data for this metric exists. Refer to the individual tab to learn more about this metric



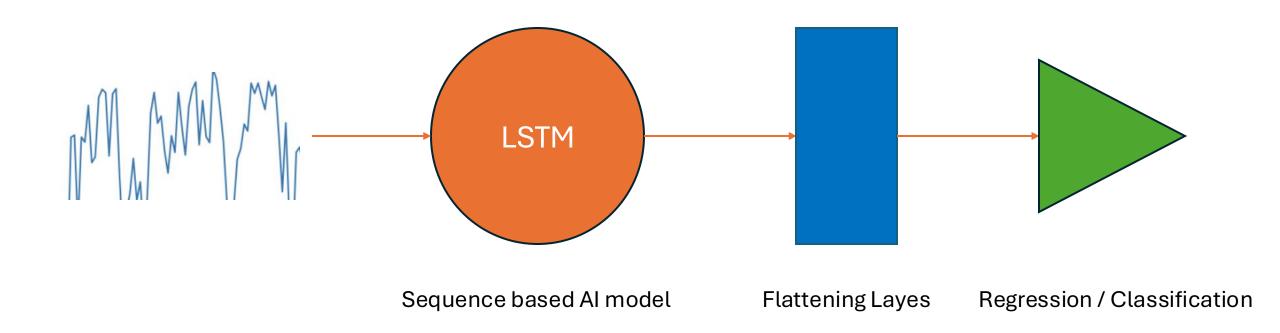
Step 5 – Data analysis pipeline

Alright time to figure out what this data means.

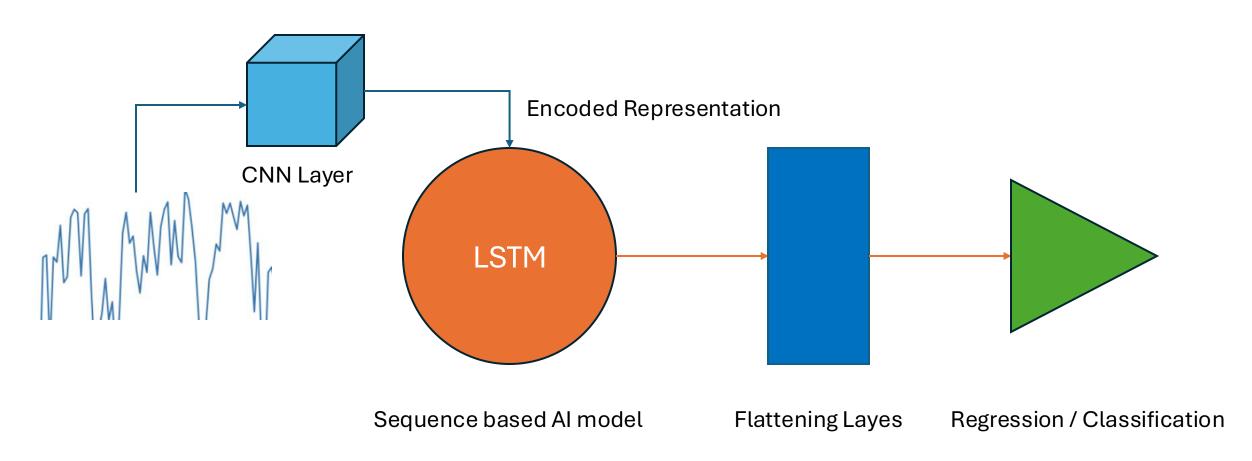
Creating an Analytic Workflow



A Simple Template to Try - 1



A Simple Template to Try - 2



Thank You Let's Stay in Touch!

Reach out at –
 shravanaras@arizona.edu

