ICON

Digital Health, Statistical Challenges and Estimands

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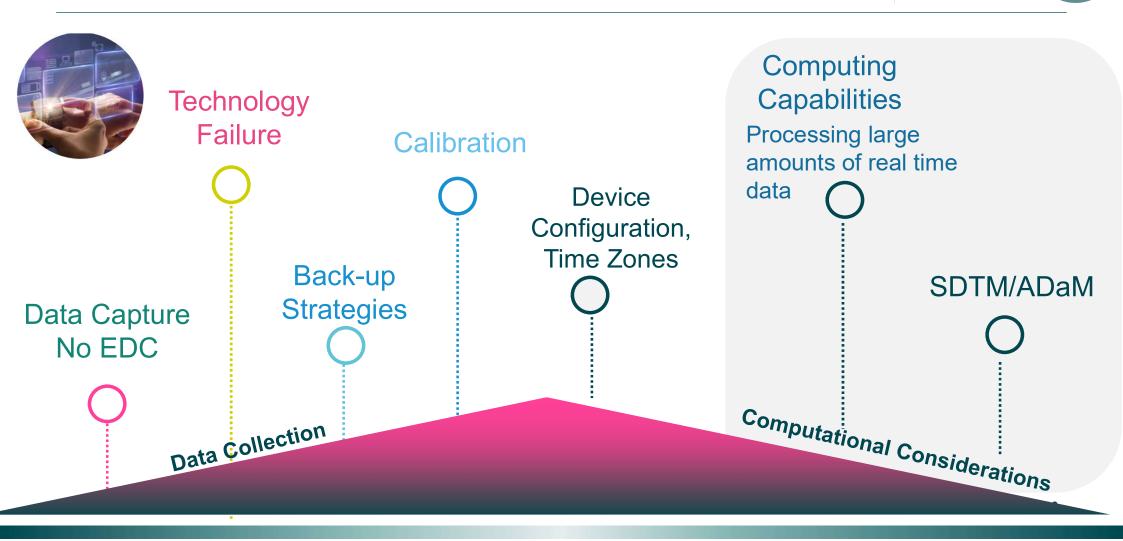
PSI, 18th June 2024



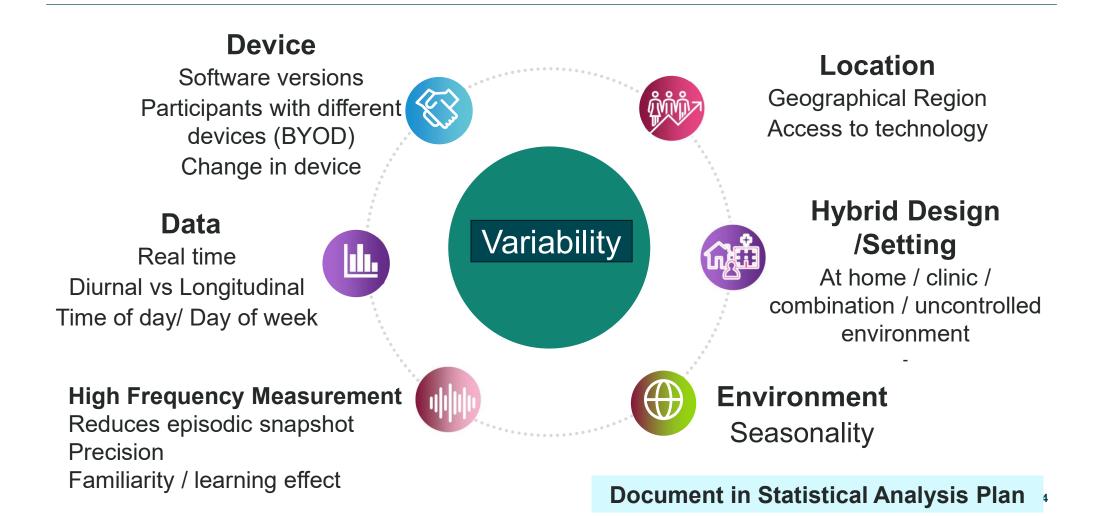
Overview

- Introduction
 - Challenges Technology
 - Challenges Study Design, Endpoints, Data
- Intercurrent Events and Estimands
- Handling Missing Data
- References

Statistical Challenges, Technology



Digital Endpoints, Variability



Study Design and Digital Endpoints

Digital Health provides the opportunity for **novel** study designs and endpoints

Study Design

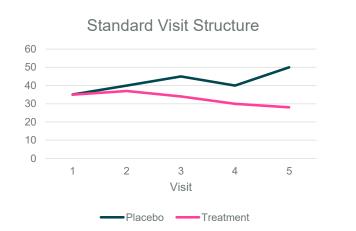
- Choice of **setting** (hybrid collection)
- Focus data collection on necessary / sufficient
- User acceptability of technology/ burden
- Ethical considerations
- Blinding
- Estimands Framework

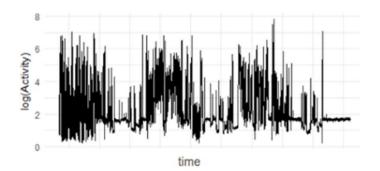
Endpoints

Variability Range Validated endpoints Regulatory approval for Primary/Secondary Effect Size/Power Meaningful change

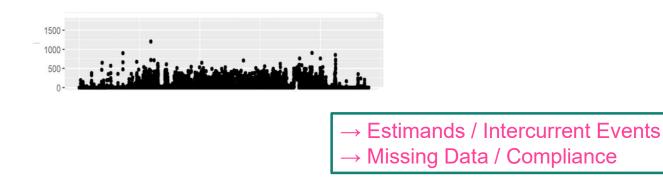


Statistical Challenges – Data and Analysis





- Variability, diurnal vs longitudinal
- Innovative methods for data analysis of time series
- Beyond reliance on change from baseline
- Generalised Additive Models, Functional Data Analysis, Change point detection
- Granularity of Data (raw vs summaries)
- Proprietary Algorithms
- AI/Machine Learning for meaningful biomarkers/ prediction



Lisi & Abellan, 2023; Karas et al, 2022; Beaulieu & Killick, 2018

Estimands and Digital Health?

Population Treatment/ Intervention Endpoint

Estimato

Intercurrent Events

Summary Measure

Estimands Framework: Inform study design and strategic thinking Can clarify added value, potential risks and opportunities for novel estimands of DCT

What: Impact of DCT on each estimand attribute?

How: DCT impacts on estimator?

Izem et al, 2024

Digital Health and Intercurrent Events

Intercurrent Events

Post baseline events that impact the presence or interpretation of endpoints

Common ICEs (Standard Trials)

- Treatment Discontinuation / Treatment Non-Adherence
- Rescue Medication
- Prohibited Medication
- Death

Specific to Digital Collection

- Frequency / likelihood /timing of identifying ICEs
- Change in Device
- Network interruption?

Missing Data



Likely to be common

Classification informative, non-informative

- Technology related (internet connectivity, battery power, device malfunction)
- Person related (compliance, time in study, disease stage, symptom severity, activity level)
- Engagement
 - how invested is the investigator
- Compliance checks and technical support / notifications.

Missing data is likely to be missing not at random

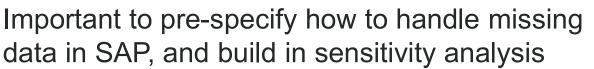
Junrui et al, 2022

Missing Data Example - Wearables

Common approach: define a **day** as missing or observed using a threshold on wear time.

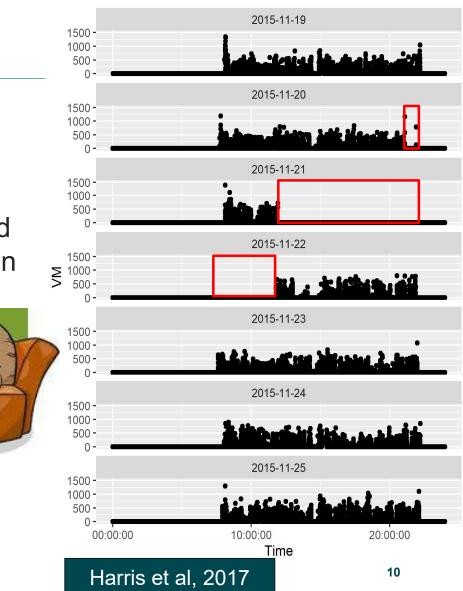
Using epoch-level data, intervals can be classified as inactivity, sleep or missing. Missing intervals can be handled with Multiple Imputation.

Most likely to be missing not at random.



data in SAP, and build in sensitivity analysis

Tackney et al, 2021, 2023



References

ICON White Paper: https://www.iconplc.com/insights/digital-disruption/mHealth-wearables/digital-endpoints/

ICON Digital Platform: https://icon.widen.net/s/ldrpdx5ppm

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