

# Assessing Uniformity in Sampling of Sedentary Times

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JSM 2021 Novel Methodology for the Analysis of Physical Activity Data Measured by  
Accelerometers

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- 1 Micro Randomized Trials (MRT), Sedentary Times and Sampling Algorithm
- 2 Assessing Uniformity of the Sampling Algorithm
- 3 Discussion of Open Problems

# A Key Question to Mobile Intervention System

- Mobile intervention: Change one's behaviour
- Trade-off in number of interventions

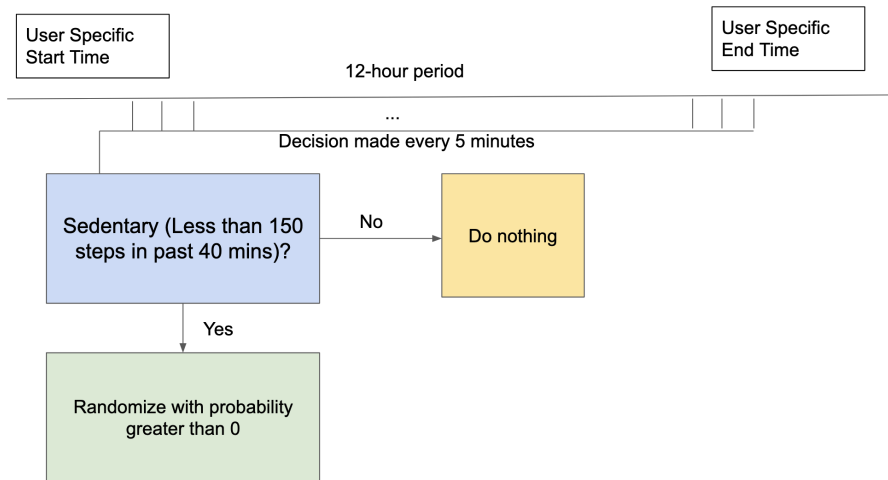
**No Enough Intervention:  
Not Effective**

**Too Much Intervention:  
User Burden**



- Context also matters.
- Key Question: How often, and in what context should a mobile intervention be provided?
- One way to address the key question is Micro-Randomized Trial (MRT).
- Our task: To design a MRT to address the key question.

# Example: MRT to Reduce Sedentary Behavior



# A Sampling Algorithm of Just-in-time Intervention is Critical

- Want to design a MRT to learn when to intervene, specifically at sedentary times.
- A sampling algorithm should spread intervention uniformly across all sedentary times.
- Uniformity
  - Uniform number of interventions per day
  - Uniform distribution on sedentary times
- To achieve this
  - Set a hard constraint on number of intervention per day
  - At each sedentary time, predict the remaining sedentary times

# HeartSteps Study

- Aim: Develop a mobile activity assistant for individuals who are newly diagnosed with Stage 1 Hypertension
- Intervention: A push notification to encourage exercise.
- Decision time: Every 5 mins
- Notation:
  - $t = 1, 2, 3, \dots$ , decision time.  $\mathcal{T}_d$  set of decision times for day  $d$ ;
  - Each day is separated into three 4-hour blocks. Denote  $\mathcal{B}_{d,k}$  as  $k$ th block on day  $d$ ,  $\bigcup_{k=1}^3 \mathcal{B}_{d,k} = \mathcal{T}_d$
  - $A_t$ : intervention;  $H_t$ : history;  $I_t$ : sedentary indicator

# Description of the Sampling Algorithm, i.e. Randomisation Probabilities

- Goal:
  - ① An average of 0.5 Intervention per Block  $\mathcal{B}_{d,k}$
  - ② Uniform distribution in sedentary times
- For Goal 1: Subtract all previous probabilities from the hard constraint.  $N_1^* - \sum_{s \in \mathcal{B}_{d,k}, s \leq t-1} I(I_s = 1)\pi_s$
- For Goal 2: If remaining sedentary times  $g_t$  is known, by setting  $N_1^* = 0.5$  we can sample uniformly with probability

$$\frac{N_1^* - \sum_{s \in \mathcal{B}_{d,k}, s \leq t-1} I(I_s = 1)\pi_s}{1 + g_t}$$

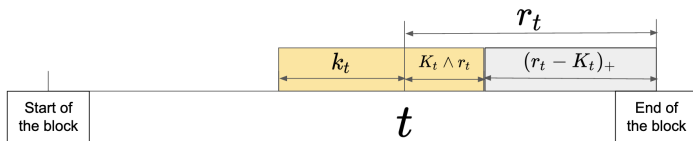
- Truncate to  $[0.005, 0.995]$

Tuning parameters:

- $g_t$ : forecast of the remaining sedentary times in the block.
- $N_1^*$ : desired average number of interventions per block.



# Model for Prediction of the Remaining Sedentary Times $g_t$



- Notation:

- $k_t$ : run length.
- $K_t$ : remaining run length of the current run
- $r_t$ : remaining time in the current block
- $F_t$ : the fraction of sedentary times outside the run.

- Estimate for  $g_t$ :

$$\hat{g}_t = \hat{E}[K_t \wedge r_t] + \hat{E}[F_t] \hat{E}[(r_t - K_t)_+]$$

- $\hat{E}[K_t \wedge r_t], \hat{E}[F_t], \hat{E}[(r_t - K_t)_+]$  are non-parametric averages from prior data.

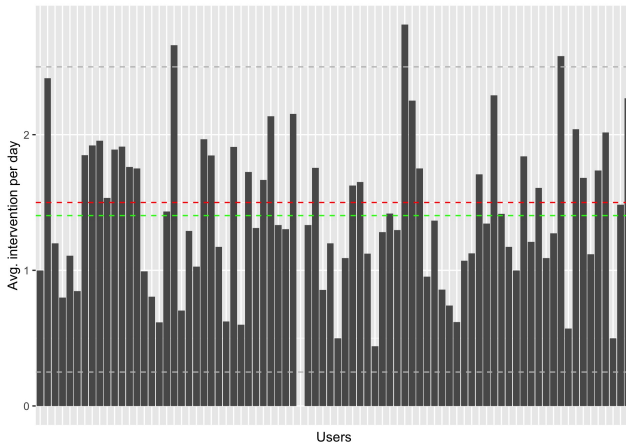
## How many interventions were delivered on each block and each day?

- We first calculate average interventions per user per block over all days.
- Then for each block, we do a summary for all users. Total 81 users are considered.
- Goal: 0.5 per block and 1.5 per day.

	Median	Mean	Std. Dev.
Block 1	0.476	0.471	0.219
Block 2	0.451	0.456	0.241
Block 3	0.448	0.476	0.259
Daily	1.333	1.404	0.568

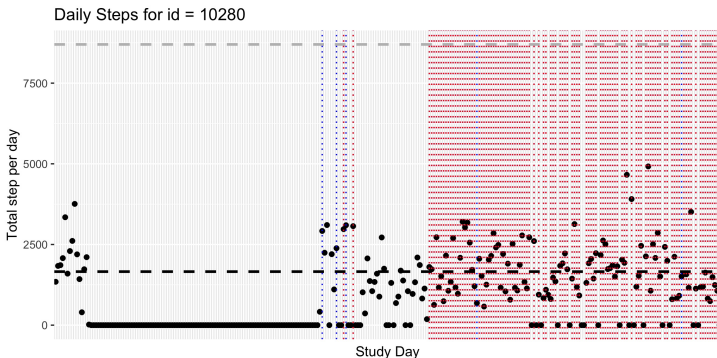
- Algorithm is good at targeting the goal.
- Std. Dev. increases with block because number of sedentary times decrease with block.

# Average interventions sent per day across users



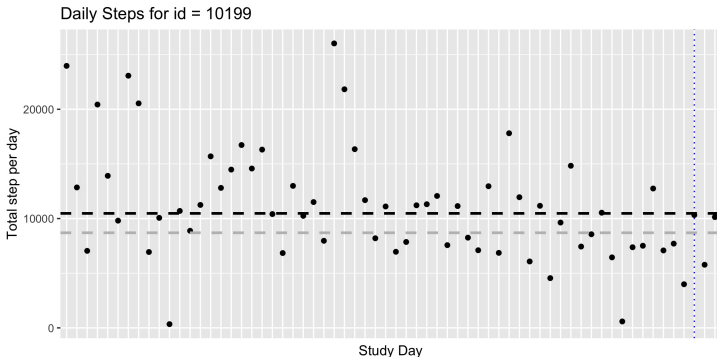
- Red line is the target (1.5 per day)
- Green line is average over users (1.404 per day).
- Grey lines: high (2.5 per day) and low (0.25 per day).

# Three Users with daily avg intervention greater than 2.5



- Each point is daily sum of Steps for the User.
- Black horizontal line: average daily steps for **this user**.
- Gray horizontal line: average daily steps across **all (user,day)**.
- Blue vertical line: at least one sedentary time on the day.
- Red vertical line: at least one intervention on the day
- Other two users have the same pattern.

# One user with daily avg intervention lower than 0.25



- Each point is daily sum of Steps for the User.
- Black horizontal line: average daily steps for **this user**.
- Gray horizontal line: average daily steps across **all (user,day)**.
- Blue vertical line: at least one sedentary time on the day.
- Red vertical line: at least one intervention on the day

# Is the Algorithm Uniform Across all Sedentary Times?

- Notation:  $N_{u,d,k}$  as the number of sedentary time per (user, day, 4-hour block).
- Measure: KL divergence **between the actual** treatment probability  $p_a$  and the **uniform** distribution defined by  $p_u = 0.5/N_{u,d,k}$ .
- We calculate mean KL divergence on each (user, day, 4-hour block) by

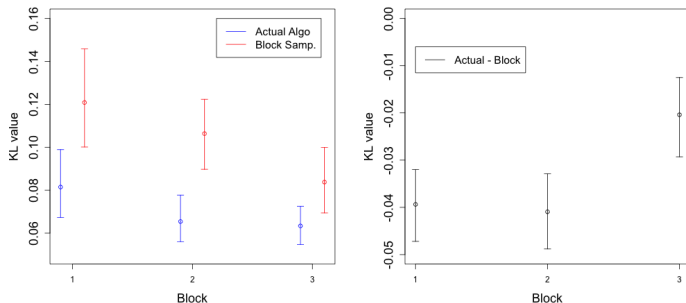
$$\frac{1}{N_{u,d,k}} \sum_{t=1}^{N_{u,d,k}} KL(p_{a,t}, p_{u,t})$$

where  $KL(p, q) = p \log(p/q) + (1 - p) \log((1 - p)/(1 - q))$

# A Comparison Using KL Divergence

A baseline algorithm—Block sampling

- First, from prior data, calculate average number of sedentary times across all (user, day) in block  $k$ :  $\hat{M}_k, k = 1, 2, 3$ .
- Then set the randomisation probability to be  $0.5/\hat{M}_k$  for sedentary time in block  $k$ .     **M1: 0.0280, M2: 0.0255, M3: 0.0256**



**Figure:** Left: KL between the actual probability(blue)/block sampling(red) and the uniform probability. Right: CI of their difference

## Discussion: How generalizable is our analysis?

- Missing sedentary times could lead to bias.
- Type A: Missing sedentary time due to bad data propagation.
- Type B: Missing sedentary time due to user behaviour.
- User population will differ in real life
  - Our analysis still gives strong indication if underlying dynamics of missingness are similar.



# Open Problems

- Other ways to measure whether the algorithm is uniform?
- How to model sedentary time such that it can form a good test bed to test algorithms?
- How would one design an algorithm with better performance?

Thanks for your attention!

Any questions?