



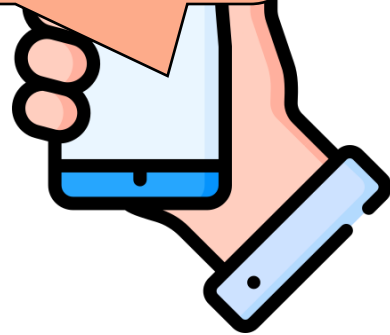
# **RoME: A Robust Mixed-Effects Bandit Algorithm for Optimizing Mobile Health Interventions**

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# Bandits in Mobile Health

Feeling low on energy?  
Energize your day with  
a short walk outside!



$t = 1$   
 $S_1, A_1, R_1$



$t = 2$   
 $S_2, A_2, R_2$



$t = 3$   
 $S_3, A_3, R_3$



$t = 1$   
 $S_1, A_1, R_1$



$t = 1$   
 $S_1, A_1, R_1$



$t = 2$   
 $S_2, A_2, R_2$



Time

# Challenges of Mobile Health Setting

- **Heterogeneity** in advantage function (treatment effects) over time and across participants
- **Nonstationary** rewards even under no treatment
- **Nonlinear** relationships between contexts and rewards
- **Network structure** among participants, such as students within a school

# Our Solution

Advantage Function:  $\mathbf{s}_{it}^T (\boldsymbol{\theta} + \boldsymbol{\theta}_i^{\text{user}} + \boldsymbol{\theta}_t^{\text{time}})$

Fixed effect

Participant  
random effect

Time random  
effect

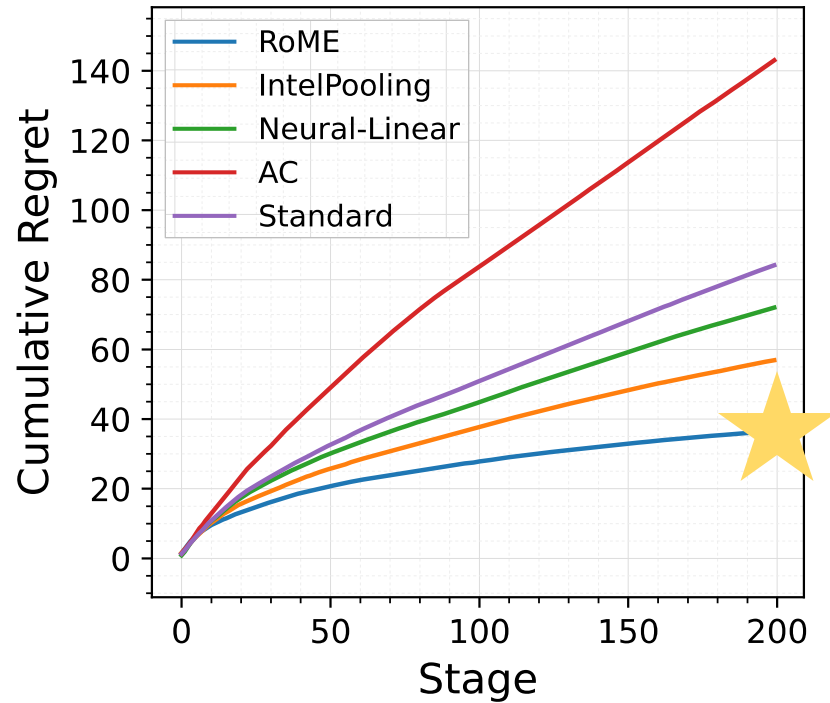
## Other Method Components

- Debiased machine learning (Chernozhukov et al. 2018)
- Nearest-neighbor regularization (Yang, Toni, & Dong 2020)
- Thompson sampling (Agrawal & Goyal 2012)

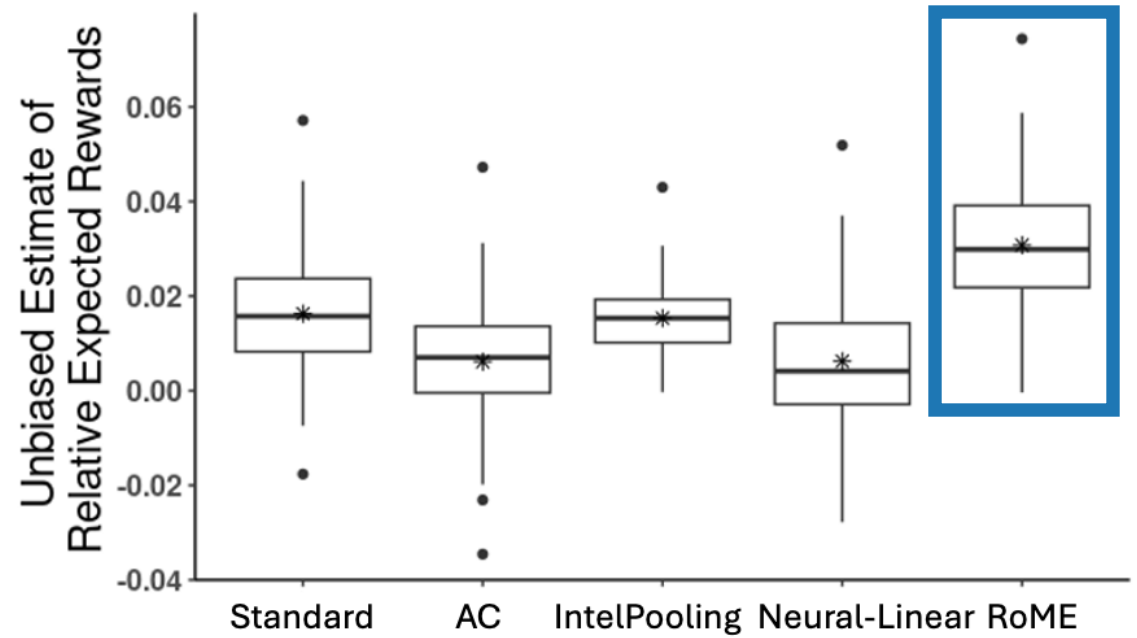
Up to log factors, the regret scales like  $\sqrt{K}$ , where  $K$  is the number of stages (time points)

# Results

## Simulation



## Off-policy Evaluation Study



Lower regret (higher rewards) across both settings!

# Thank you!

Check out our repository at [github.com/eastonhuch/RoME](https://github.com/eastonhuch/RoME).

Contact me with questions at [ekhuch@umich.edu](mailto:ekhuch@umich.edu).

# References

- Agrawal, Shipra, and Navin Goyal. Analysis of Thompson sampling for the multi-armed bandit problem. In *Conference on Learning Theory*, pp. 1-39. JMLR Workshop and Conference Proceedings, 2012.
- Chernozhukov, V., Chetverikov, D., Demirer, M., Duflo, E., Hansen, C., Newey, W., and Robins, J. Double/debiased machine learning for treatment and structural parameters. *The Econometrics Journal*, 21(1):C1–C68, 01 2018. ISSN 1368-4221. doi: 10.1111/ectj.12097. URL <https://doi.org/10.1111/ectj.12097>.
- Yang, K., Toni, L., and Dong, X. Laplacian-regularized graph bandits: Algorithms and theoretical analysis. In *International Conference on Artificial Intelligence and Statistics*, pp. 3133–3143. PMLR, 2020.