

Introduction

Boosting cohort engagement in mHealth programs

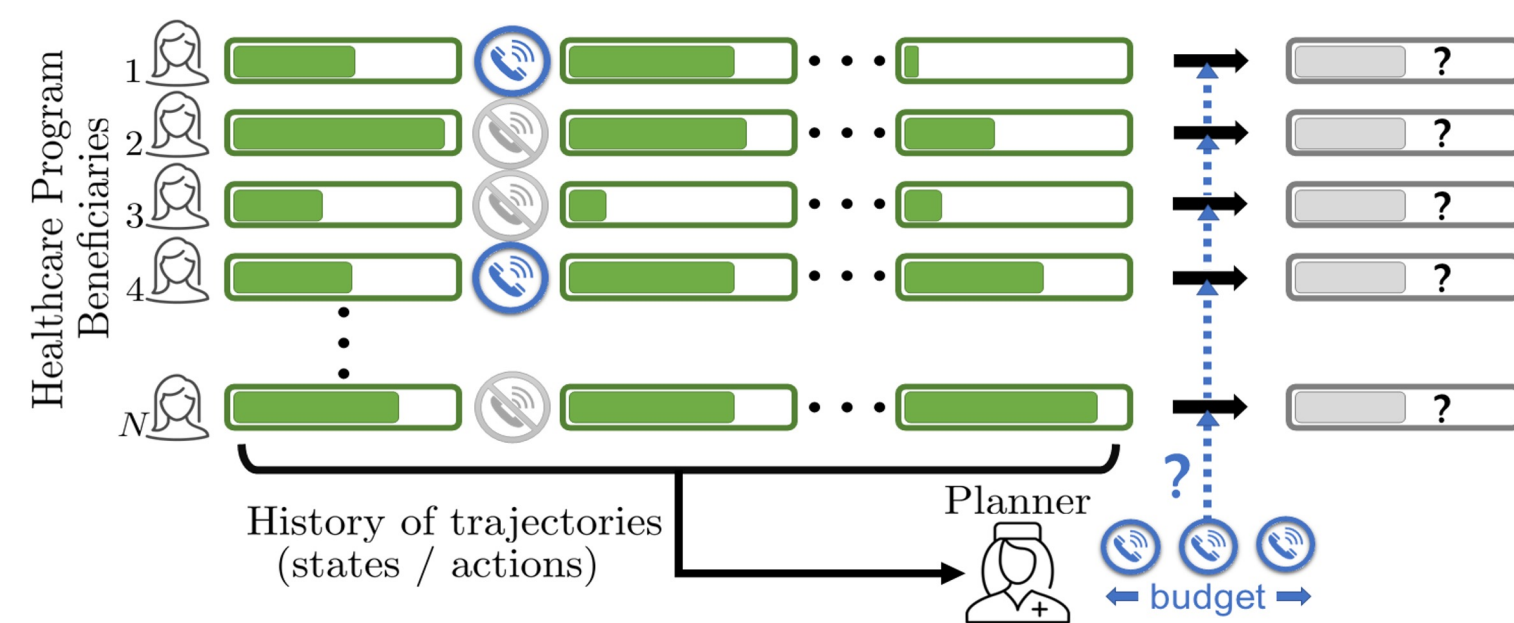
Key Challenges:

1. Scheduling limited intervention resources in sequential planning problem
2. Inferring preferred time slot for automated voice messages

CHAHAK: Campaign for Health-resource Allocation using Historical data in AI based system for Kilkari

- Model Multi-Action Time Series Bandit
- First work optimising time slot section

Human behavior is likely to contain **temporal dependencies**



Kilkari: India's Ministry of Health & Family Welfare mHealth program

World's largest maternam mHealth program



- Maternal Health
- Immunization
- Child health
- Family Planning

automated voice messages sent for **72 weeks** to **3 million** active subscribers

Mission: Reduce maternal, neonatal and child mortality and morbidity in underprivileged communities

Results

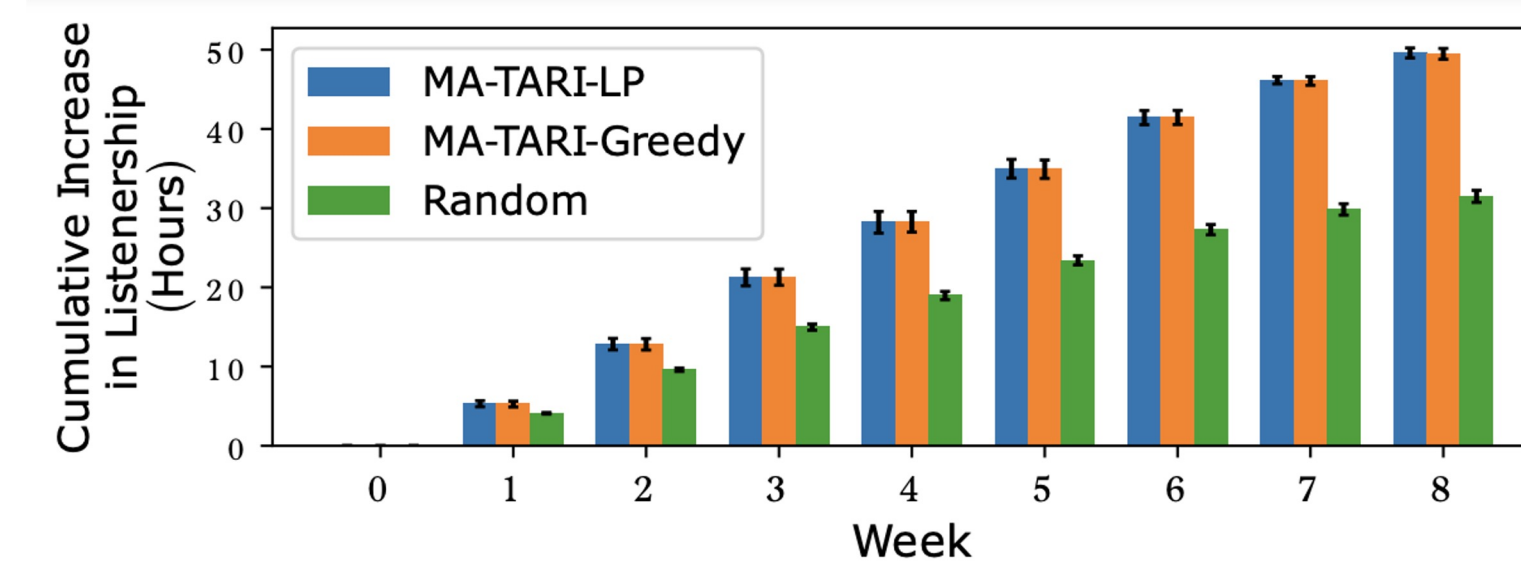
Experiment Setup - Intervention Planning

We consider the following methods:

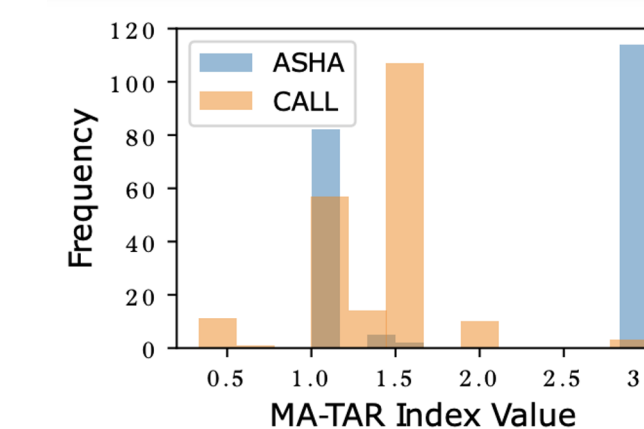
1. MA-TARI Integer Linear Programming
2. MA-TARI Greedy
3. Random - randomly allocate interventions
4. Control - No interventions

Simulated experiment: Consider 4000 beneficiaries in simulation and allow 1% ASHA interventions and 1% CALL interventions every week

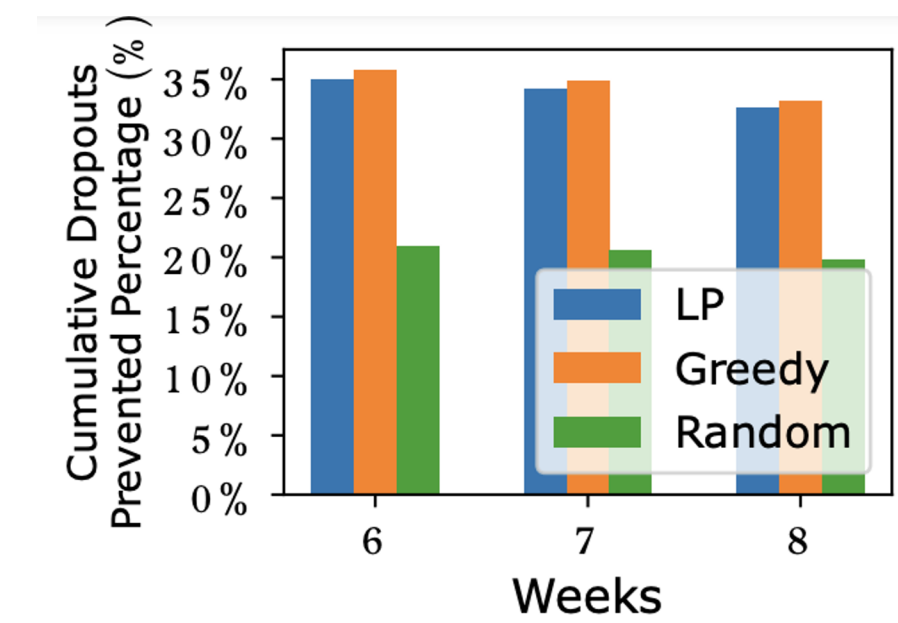
Results - Intervention Planning



Real world significance: 49.6h and 49.47h of additional content listened by the beneficiaries by the 8th week (for the LP and Greedy variants, respectively). **57% increase in exposure** compared to Random

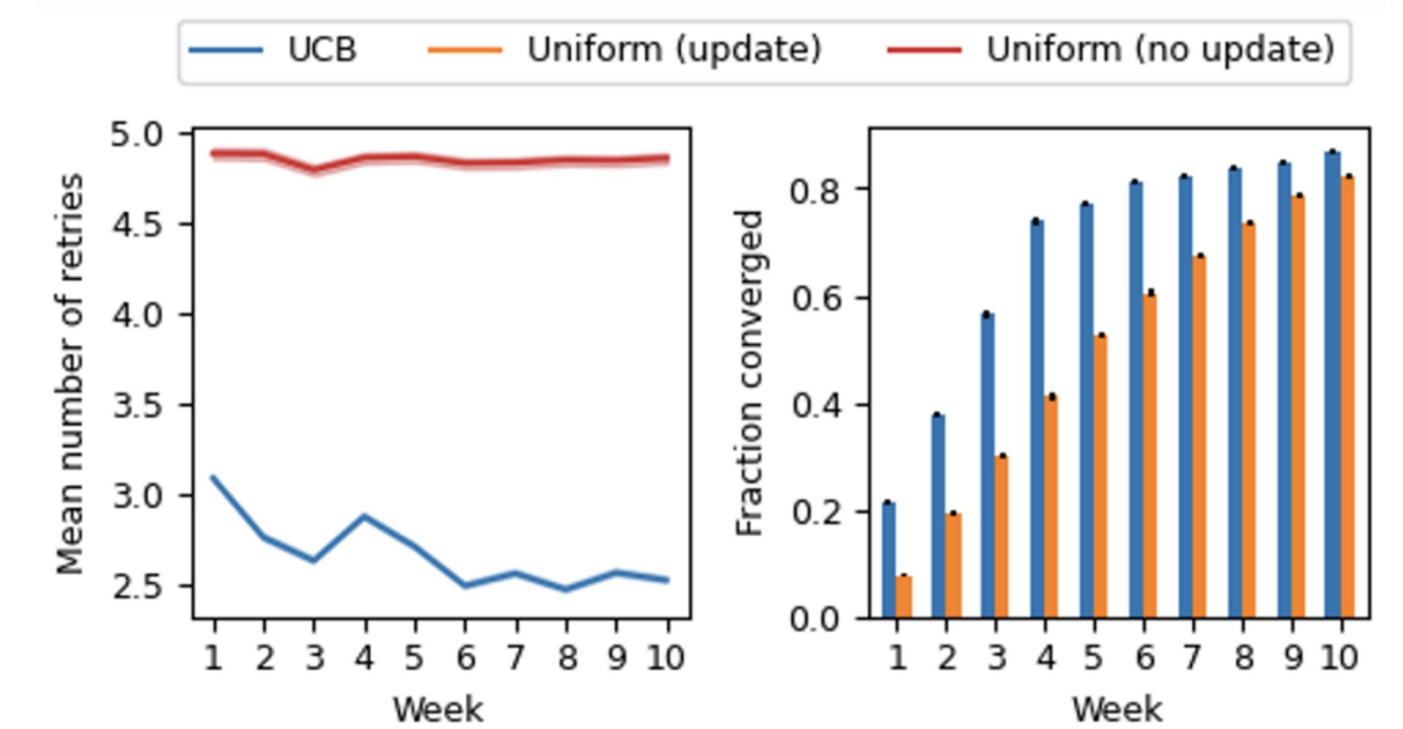


Potentially delay the poor listenership state by 3x weeks



MA-TARI cumulatively prevents **33% more dropouts** by 8th week as compared to Random

Results - Time Slot Planning



[LEFT] *Uniform (no update)* requires an average of **4.86 calls per week**, while *UCB* requires **2.5 calls per week** by week 6

[RIGHT] *UCB* achieves a pickup rate difference of **less than 0.15**, compared to the true pickup rate **within 4 weeks**, for 75% of the population, compared to **8 weeks** by *Uniform (update)*

Method

Problem Domain: Intervention Planning Under Limited Resources

Scheduling limited multiple interventions to maximize engagement in a mobile health program

Model & Objective 1: Beneficiary as arms of bandit model[1], maximize # of beneficiaries in engaging state.

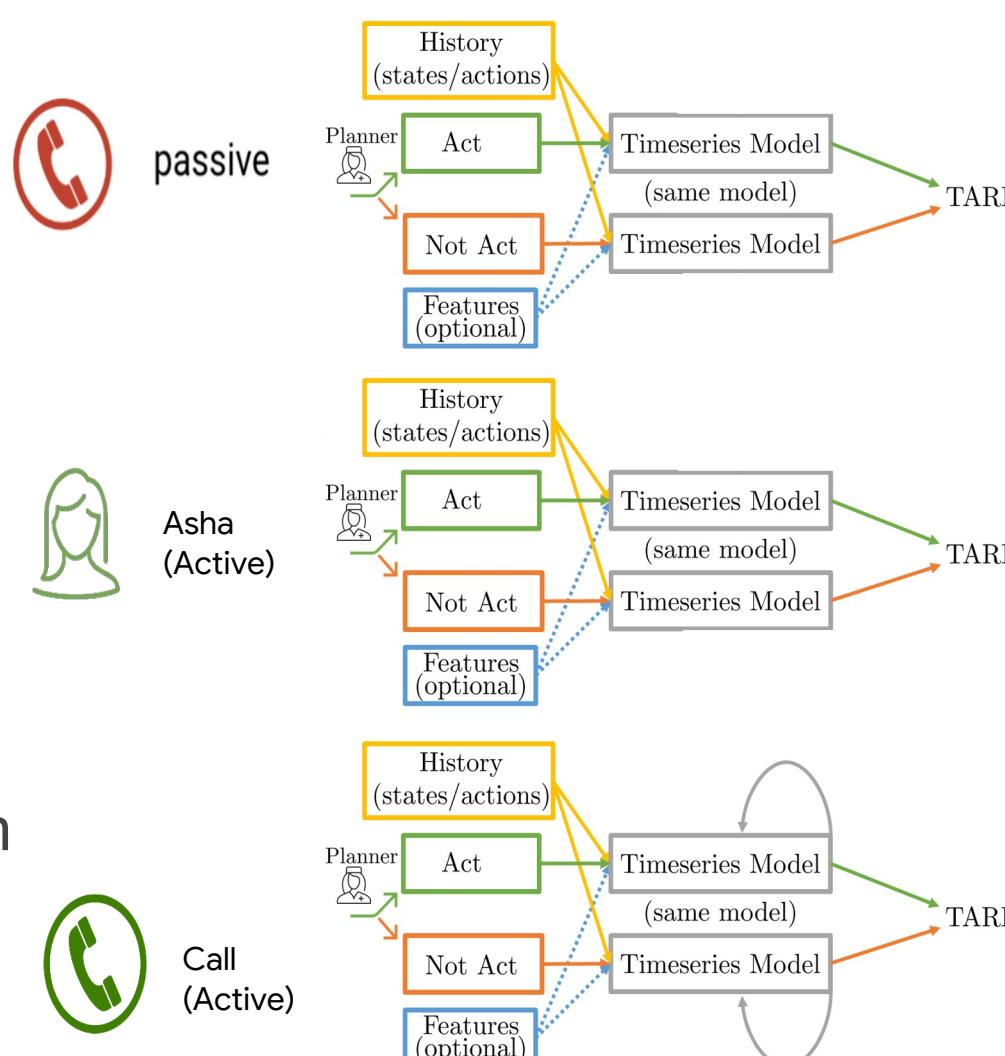
Solution: Time Series Bandits (TSBs) and Multi Action TAR [2] approach using Greedy and LP selection strategies

Model & Objective 2: Time slots as arms of bandit model, infer preferred time slot of every beneficiary for receiving automated voice messages.

Solution: Run a Upper Confidence Bound (UCB) model for each each beneficiary. The timeslot (a_τ at timestep τ) is chosen based on the following UCB equation:

$$a_\tau = \arg \max_j \mu_j + \sqrt{2(\log \tau) / \nu_j}$$

Multi Action - Time-series Arm Ranking Index (MA-TARI)



Conclusion

1. Proposed **Multi-Action Time-series Arm Ranking Index (MA-TARI)**: (i) act on beneficiaries which will benefit the most from an intervention (ii) use greedy selection strategy for multiple actions - practical and scalable (iii) 49hrs more content listened (iv) 57% increase in engagement and 33% more dropouts prevented compared to Random
2. Propose **Upper Confidence Bound** based strategy for Call Time Slot optimization: 47% reduction in call attempts rate by week 6

Path to Deployment

1. Nationwide deployment of CHAHAK through collaboration with our NGO partner ARMMAN
2. Planned cohort study to validate CHAHAK's benefits on ground
3. Draw from experience of deploying SAHELI to tackle the challenges that come up

References

1. Field Study in Deploying Restless Multi-Armed Bandits: Assisting Non-profits in Improving Maternal and Child Health. Mate et al., AAAI 2022
2. Limited Resource Allocation in a Non-Markovian World: The Case of Maternal and Child Healthcare. Danassis et al., IJCAI 2023