IE 613 Project Presentation

Multi-player multi-armed Stochastic Bandits for Energy Harvesting

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Introduction

What is Energy Harvesting

- Energy Harvesting is the process of converting energy that is obtained from external energy sources such as wind, thermal, kinetic and solar energy into electricity.
- RF energy is one form of electromagnetic energy which consists of radio waves of magnetic and electrical energy radiating through free space. RF-EH becomes a propitious solution for wireless networks with a limited lifetime

Why is Energy Harvesting becoming a buzz word?

- The market has noticed a tremendous growth in the energy consumption patterns due to continuously expanding network of connected devices
- The alternative wireless information and power transfer techniques are important from the operational cost saving apart from the theoretical research.
- It allows wireless nodes to recharge their batteries from the radio frequency signals instead of fixed power grids and traditional sources of energy

Technologies used in Harvesting Energy

 Wireless Power Transfer (WPT): uses the transmission of electrical energy from a power source, using electromagnetic fields to the electrical component in the circuit that uses electrical power without wired connections.

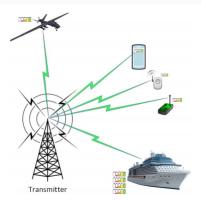
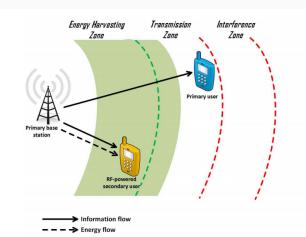


Fig. 2. RF Wireless Power Transfer - Devices such as cellular phone, sensors, drones, electronic devices in vessels receive and harvest energy from RF signals and recharge their batteries.

Technologies used in Harvesting Energy

Simultaneous Wireless Information and Power Transfer (SWIPT): It
enables the simultaneous transfer of information and power wirelessly
and hence a trade-off has to be made between the information rate
and the harvested energy level. This trade-off varies with the system.



Technologies used in Harvesting Energy

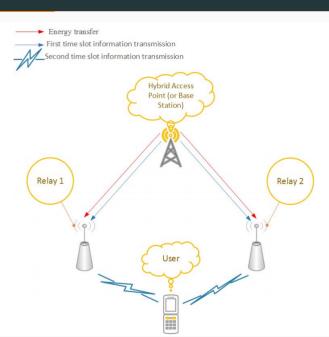
 SWIPT assisted Device to Device Communication: It provides a direct connection between the users with partial or almost no network infrastructure utilization.



Efficient ways of Energy Harvesting

Cooperate Relaying: In a wireless communication network, the
location of nodes vary and sometimes even hinder each other nodes
due to geographical or climatic reasons, leading to infeasible
communication in the line of sight.Relay nodes acts as such
intermediate idle nodes which overhears the source's transmission
and re-transmit its modified version to the destination to help the
overall decoding process.

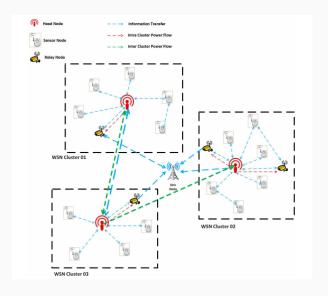
Relay Network



Efficient ways of Energy Harvesting

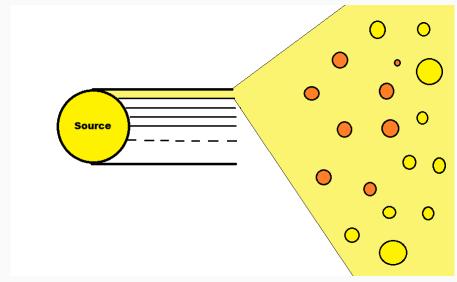
 Clustering: clustering the wireless sensor networks together with applying SWIPT to recharge low energy relay nodes. In such a setting, we would consider a wireless sensor network consisting of multiple clusters of sensors and a common sink node. This sink node is responsible for collecting data from all the sensors. The head node of each cluster works as information and power transferrer. Relay nodes harvest the radio frequency energy from this head node of their cluster and uses this energy to recharge the batteries.

Clustering of Wireless Sensor Networks

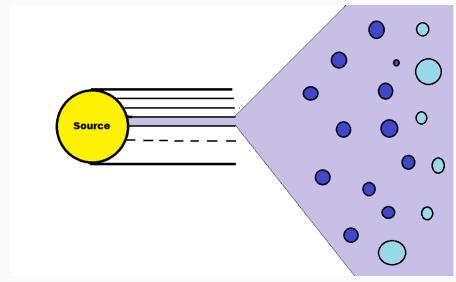


Energy Harvesting in an Online Setting

Propagation of Energy through a particular channel



Energy received differentially with different channels



Assumptions

- We assume that no node hinders the other in any of the K channels
- Our Source is the main Radio Frequency emitting source
- We are not assuming any sort of clustering of the nodes or sources
- The models are in an energy harvesting setting only and does not include Information transfer simultaneously or otherwise

Approaching the Problem statement

 We might want to know the channel which maximises the total energy consumption across all energy harvestors. i.e. we would like to find:

$$\arg \max_{k} \ \sum_{n \in [N]} \mu_{kn}$$

In the previous approach to the problem, it might so happen that
the optima is achieved for a value of k for which one or more nodes
might die out. In that case, we can frame our objective function as
the maximum energy consumed across all nodes such that none of
the nodes die out. i.e. we would like to find:

$$\underset{k}{\operatorname{arg }} \max_{k} \min_{n} \mu_{nk}$$

Input: No. of channels K; No. of nodes N.

Step 1: for t=1,2,...,K, do:

Source emits energy through channel t. Observe the energy harvested by each node and calculate the estimated gain of all nodes $\hat{\mu_t}$ for channel t.

Step 2: for t=K+1, K+2, ..., do:

$$K_t = arg \max_k \{\hat{\mu_k} + \sqrt{rac{lpha \log(t)}{ ext{Total No. of times channel k has been used}}}\}$$

Source emits energy through channel K_t . Observe the energy harvested by each node and calculate the estimated gain of all nodes $\hat{\mu_{K_t}}$ for channel t.

Step 3: At whichever round we decide to stop, the channel K_t found at that last round would give us the optimum value to our problem. where

$$\hat{\mu_k} = \frac{\sum_{t=1}^T \sum_{n \in N} X_{nkt}}{\textit{Total No. of times channel k has been selected till T rounds}}$$

Input: No. of channels K; No. of nodes N.

Step 1: for t=1,2,...,K, do:

Source emits energy through channel t. Observe the energy harvested by each node and calculate its estimated gain $\hat{\mu_{tn}}$ for channel t.

Step 2: for t=K+1,K+2, ..., do:

for each channel k, compute n_k as:

$$n_k = \arg\min_n \{ \hat{\mu_{kn}} - \sqrt{\frac{\alpha \log(t)}{\textit{Total No. of times channel k has been used}} \}$$

Now, determine k_t^* as:

$$k_t^* = arg \max_k \{ \hat{\mu_{kn_k}} - \sqrt{\frac{\alpha \log(t)}{Total \ No. \ of \ times \ channel \ i \ has \ been \ used} }$$

Source emits energy through channel k_t^* . We observe the energy harvested by each node and calculate the estimated gain of them $\hat{\mu_{K_t}}$ for channel t.

Step 3: At whichever round we decide to stop, the channel k_t^* found at that last round would give us the optimum value to our problem.

Analyzing the Proposed

Algorithms

We conducted the simulation for K=10, N=50, T=25000 and sample paths=20. We also fixed the parameters μ_{nk} as $\mu_{nk} = \frac{nk}{(K+1)(N+1)^2}$ for n=1(1)50 and k=1(1)10.

We considered two different cases:

- Gains follow Bernoulli distribution with the above parameters.
- Gains follow Exponential distribution with the above parameters.

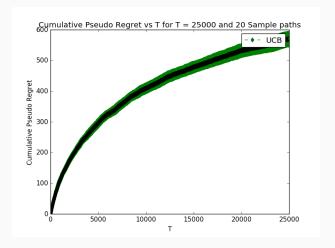


Figure 1: Pseudo Cumulative Regret vs T [Gains∼ Benoulli]

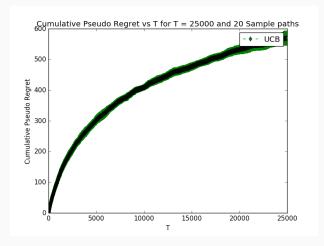


Figure 2: Pseudo Cumulative Regret vs T [Gains∼ Exponential]

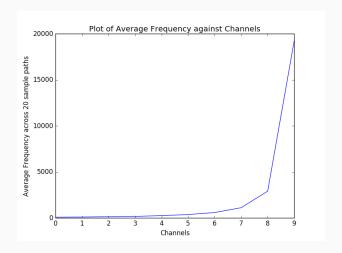


Figure 3: Avg. Frequency vs Channels [Gains \sim Bernoulli]

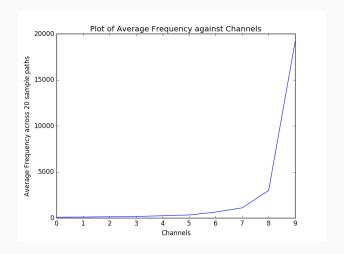


Figure 4: Avg. Frequency vs Channels [Gains~ Exponential]

We conducted the simulation for K=8, N=15, T=25000 and sample paths=20. We also fixed the parameters μ_{nk} as $\mu_{nk} = \frac{nk}{(K+1)(N+1)^2}$ for n=1(1)15 and k=1(1)8.

We considered two different cases:

- Gains follow Bernoulli distribution with the above parameters.
- Gains follow Exponential distribution with the above parameters.

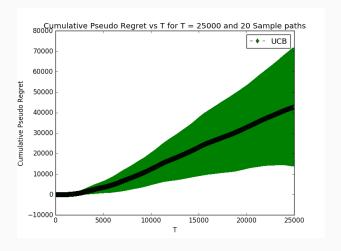


Figure 5: Pseudo Cumulative Regret vs T [Gains∼ Benoulli]

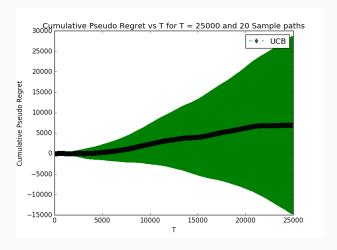


Figure 6: Pseudo Cumulative Regret vs T [Gains∼ Exponential]

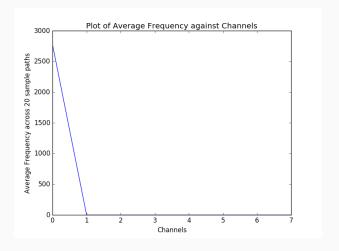


Figure 7: Avg. Frequency vs Channels [Gains \sim Bernoulli]

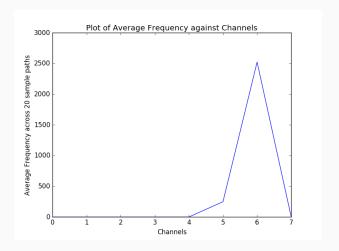


Figure 8: Avg. Frequency vs Channels [Gains~ Exponential]

Thank you