

On Outage Optimal Transmission Policies for Energy Harvesting Multi-hop Links

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Outline

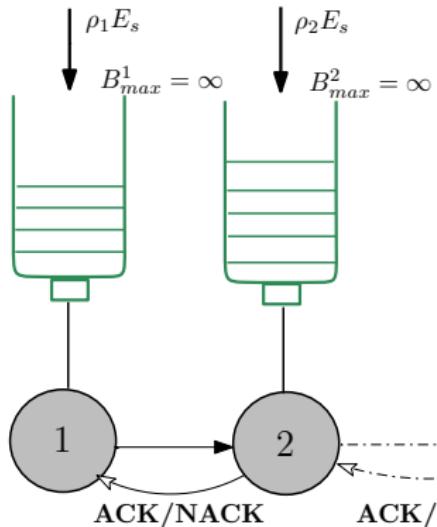
- ▶ Introduction
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Introduction

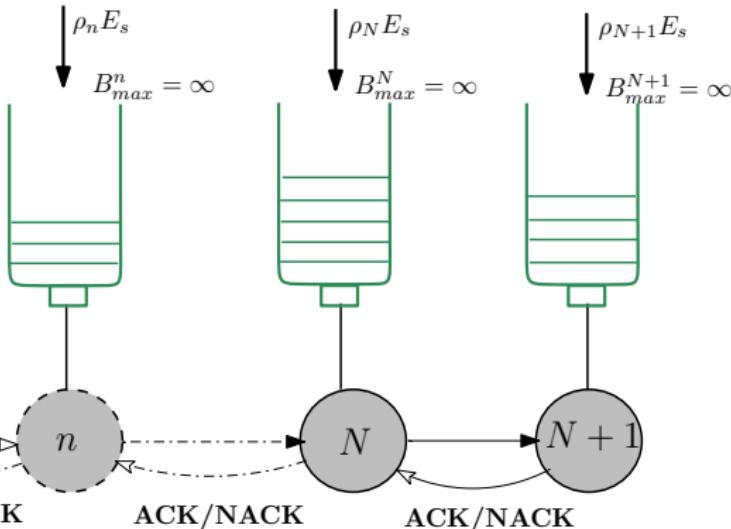
- ▶ *Sporadically* energy is harvested from environment for eg. solar, wind etc
- ▶ *Energy neutrality constraint* (ENC): cumulative energy used cannot exceed the total harvested energy
- ▶ Energy neutrality constraint: infinite number of constraints
- ▶ Central issue: design of energy management policies to optimize a utility function
- ▶ Policy: prescription of the transmit power on the basis of available system-state information

System Model

SOURCE



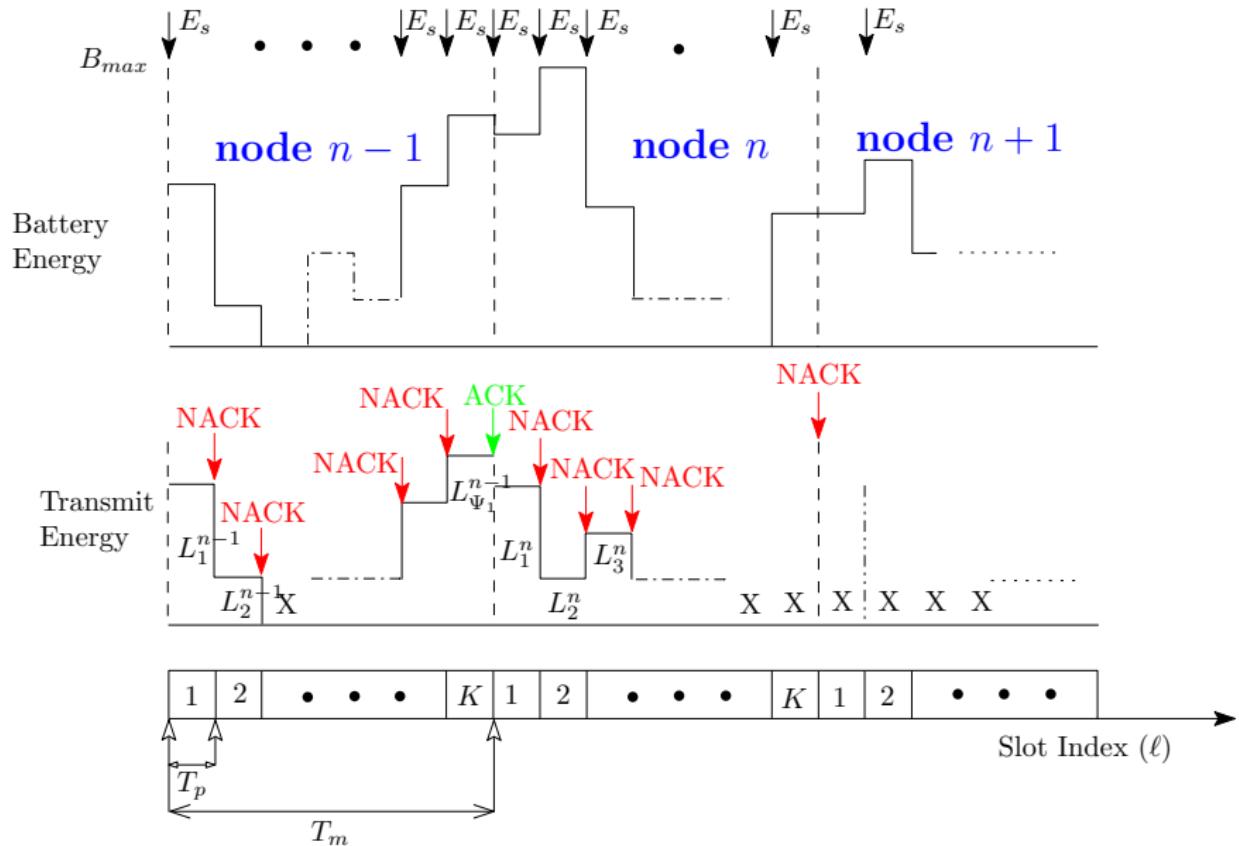
DESTINATION



System Model

- ▶ N -hop EH link with **block fading channel**
- ▶ All nodes are EH nodes (EHN)
- ▶ Periodically gets a packet, to be delivered by a deadline (multi-hop frame duration T_f)
- ▶ NK slots of duration T_p , $NK \triangleq \lfloor T_f / T_p \rfloor$
- ▶ Harvesting rate ρE_s per slot, $\forall n$
- ▶ Retransmission protocol: ARQ, HARQ-CC
- ▶ Rx sends ACK/NACK, for decoding success/failure
- ▶ Tx does not have access to CSI
- ▶ A packet remains in **outage** if not decoded correctly
- ▶ Decoding cost for each attempt = RE_s
- ▶ Packet is **dropped** if doesn't reach $N + 1^{\text{th}}$ node by the end of the frame

System Dynamics



Problem Statement

- The *drop probability*

$$P_D = 1 - \Pr[N+1]$$

$\Pr[N+1] : \Pr[N + 1^{\text{th}} \text{ node receive the packet correctly}]$

- Design goal:

$$\begin{aligned} & \min_{\{E_1^n, E_2^n, \dots, E_K^n \geq 0\}_{n=1}^N} P_D \\ \text{s.t., } & \sum_{t=1}^{t_0} \rho_n E_s \geq \sum_{t=1}^{t_0} E_t^n \quad \forall n, t_0 \end{aligned}$$

Modified Energy Neutrality Constraint

- ▶ Average Power Constraint (APC): on average, EHN consumes energy lower than the harvesting rate
- ▶ APC with large battery capacity:
 - ▶ Battery evolution has a net positive drift
 - ▶ Battery has sufficient energy to make all K attempts
 - ▶ It is throughput optimal
- ▶ For large battery system operating under APC, ENC is *equivalent to APC*
- ▶ Infinite battery approximation
 - ▶ Maximum transmit power is limited by the RF front-end hardware
 - ▶ Finite number of attempts per packet

Modified Problem Statement: ARQ Slow Fading

$$\min_{\left\{ \left\{ E_k^n \right\}_{k=1}^{K_n} \right\}_{n=1}^N} P_{\text{out}} = 1 - \Pr [N + 1]$$

subject to

$$\begin{aligned} & \Pr [n] \cdot \sum_{k=1}^{K_n} E_k^{n*} \mathbb{E}_\gamma \left\{ \prod_{i=1}^{k-1} P_e(E_i^{n*}, \gamma) \right\} + \Pr [n-1] \cdot R E_s \cdot \sum_{k=1}^{K_{n-1}} \mathbb{1}_{\{E_k \neq 0\}} \\ & \cdot \mathbb{E}_\gamma \left\{ \prod_{i=1}^{k-1} P_e(E_i^{(n-1)*}, \gamma) \right\} \leq N K \rho_n E_s \quad \forall \quad n = 1, \dots, N+1 \end{aligned}$$

where,

$$\Pr [n] = \prod_{n'=1}^{n-1} \mathbb{E}_\gamma \left\{ \sum_{i=1}^{K_{n'}} \left(1 - P_e(E_i^{n*}, \gamma) \right) \prod_{\ell=1}^{i-1} P_e(E_\ell^{n*}, \gamma) \right\}$$

Alternative Formulation

- ▶ Subproblem 1:

$$\min_{\{(Tx_1, Rx_2), (Tx_2, Rx_3), \dots, (Tx_N, Rx_{N+1})\}} P_{\text{out}}$$

s.t.

$$Tx_1 \leq NK\rho_1 E_s$$

$$Rx_2 + Tx_2 \leq NK\rho_2 E_s$$

⋮

$$Rx_N + Tx_N \leq NK\rho_N E_s$$

$$Rx_{N+1} \leq NK\rho_{N+1} E_s$$

where, Rx_n , and Tx_n are average energy spent for transmission and reception by n^{th} node

Subproblem 2 (SP2)

For all $n = 1, \dots, N$

$$\min_{\{E_1^n, \dots, E_{K_n}^n\}} P_{\text{out}}^n = \mathbb{E}_\gamma \left\{ \prod_{\ell=1}^{K_n} P_e(E_\ell^n, \gamma) \right\}$$

s.t.

$$\sum_{k=1}^{K_n} E_k^n \mathbb{E}_\gamma \left\{ \prod_{i=1}^{k-1} P_e(E_i^n, \gamma) \right\} \leq \frac{T x_n}{\Pr[n]} = T_n$$

$$R E_s \cdot \sum_{k=1}^{K_n} \mathbb{1}_{\{E_k \neq 0\}} \mathbb{E}_\gamma \left\{ \prod_{i=1}^{k-1} P_e(E_i^n, \gamma) \right\} \leq \frac{R x_{n+1}}{\Pr[n]} = R_n$$

$$0 \leq E_k^n \leq E_{\max} \quad \forall \quad k = 1, \dots, K_n$$

Equality of Optimal Solutions

Theorem

Let $\bar{E}^* = \{\bar{E}_1^*, \dots, \bar{E}_N^*\}$, and $\bar{E}^{**} = \{\bar{E}_1^{**}, \dots, \bar{E}_N^{**}\}$ be the solutions to the original, and alternative problems, respectively, then

$$P_{out}(\bar{E}^*) = P_{out}(\bar{E}^{**})$$

Proof.



SP2 for Rayleigh fading

$$\min_{\{E_1, \dots, E_K\}} P_{\text{out}} = \left(1 + \sum_{\ell=1}^K \frac{E_\ell}{N_0} \right)^{-1}$$

s.t.

$$RE_s \cdot \left[1 + \sum_{k=2}^K \mathbb{1}_{\{E_k \neq 0\}} \left(1 + \sum_{\ell=1}^{k-1} \frac{E_\ell}{N_0} \right)^{-1} \right] \leq R_n$$

$$0 \leq E_k \leq E_{\max} \quad \forall k = 1, \dots, K$$

Geometric Programming: Terminology

- ▶ Monomial: $f : \mathbf{R}_{++}^n \rightarrow \mathbf{R}$:

$$f(\mathbf{x}) = dx_1^{a^{(1)}} x_2^{a^{(2)}} \dots x_n^{a^{(n)}}$$

where, $d \geq 0$, and $a^{(j)} \in \mathbf{R}, j = 1, \dots, n$

- ▶ Posynomial: Sum of monomials

$$f(\mathbf{x}) = \sum_{k=1}^K d_k x_1^{a_k^{(1)}} x_2^{a_k^{(2)}} \dots x_n^{a_k^{(n)}}$$

where, $d_k \geq 0$, $a_k^{(j)} \in \mathbf{R}$, $k = 1, 2, \dots, K$, $j = 1, 2, \dots, n$

- ▶ Examples:

Posynomial : $2x_1^{-\pi} x_2^{0.5} + 3x_1 x_3^{100}, \frac{x_1}{x_2}$

Not a Posynomial : $x_1 - x_2, \frac{x_1 + x_2}{x_3 + x_1}$

Geometric Programming: Standard form

Standard form: (non convex problem)

$$\begin{aligned} & \min f_0(\mathbf{x}) \\ \text{subject to } & f_i(\mathbf{x}) \leq 1 \quad \forall \quad i = 1, 2, \dots, m \\ & h_\ell(\mathbf{x}) = 1 \quad \forall \quad i = 1, 2, \dots, M \end{aligned}$$

where,

$$f(\mathbf{x}) = \sum_{k=1}^{K_i} d_{ik} x_1^{a_{ik}^{(1)}} x_2^{a_{ik}^{(2)}} \dots x_n^{a_{ik}^{(n)}}$$

$$\text{and, } h_\ell(\mathbf{x}) = d_\ell x_1^{a_\ell^{(1)}} x_2^{a_\ell^{(2)}} \dots x_n^{a_\ell^{(n)}}$$

Geometric Programming: Convex form

- Let $y_i = \log x_i$, $b_{ik} = \log d_{ik}$, and $b_\ell = \log d_\ell$

$$\min P_0(\mathbf{y}) = \log \sum_{k=1}^{K_0} \exp(\mathbf{a}_{0k}^T \mathbf{y} + b_{0k})$$

$$\text{s. t. } P_i(\mathbf{y}) = \log \sum_{k=1}^{K_i} \exp(\mathbf{a}_{ik}^T \mathbf{y} + b_{ik}) \leq 0 \quad \forall \quad i = 1, 2, \dots, m$$

$$q_\ell(\mathbf{y}) = \mathbf{a}_\ell^T \mathbf{y} + b_\ell = 0 \quad \forall \quad \ell = 1, 2, \dots, M$$

► References:

- M. Chiang, "Geometric Programming for Communication Systems," *Foundations and Trends of Communication and Information Theory*, vol. 2, no 1-2, pp 1-156, Aug. 2005
- S. Boyd, S. J. Kim, L. Vandenberghe, and A. Hassibi, "A Tutorial on Geometric Programming," *Optimization and Engineering*, pp 67-127, April 2007.

SP2 as a Geometric Program

Let $z_i = 1 + \frac{1}{\mathcal{N}_0} \sum_{\ell=1}^i E_\ell \implies E_i = (z_i - z_{i-1})\mathcal{N}_0$, and $Z_0 = 1$

$$\max_{\{z_1, \dots, z_K\}} (z_K - 1)$$

s.t.

$$\sum_{\ell=1}^K z_\ell z_{\ell-1}^{-1} \leq \frac{T_n}{\mathcal{N}_0} + K$$

$$\sum_{k=1}^{K-1} \mathbb{1}_{\{z_{i+1} \neq z_i\}} z_i^{-1} \leq \frac{R_n}{RE_s} - 1$$

$$0 \leq (z_\ell - z_{\ell-1})\mathcal{N}_0 \leq E_{\max} \quad \forall \ell = 1, \dots, K$$

Approximation for posynomial ratio

- Let $g(\mathbf{x}) = \sum_i u_i(\mathbf{x})$. Approximate a ratio of polynomials $\frac{f(\mathbf{x})}{g(\mathbf{x})}$ with $\frac{f(\mathbf{x})}{\tilde{g}(\mathbf{x})}$ where

$$\tilde{g}(\mathbf{x}) = \prod_i \left(\frac{u_i(\mathbf{x})}{\alpha_i} \right)^{\alpha_i} \leq g(\mathbf{x})$$

- Directly follows from AM-GM inequality $\sum_i \alpha_i v_i \geq \prod_i v_i^{\alpha_i}$
- If, $\alpha_i = \frac{u_i(\mathbf{x}_0)}{g(\mathbf{x}_0)}$ $\forall i$, for any fixed $\mathbf{x}_0 > 0$, then

$$\tilde{g}(\mathbf{x}_0) = g(\mathbf{x}_0)$$

- $\tilde{g}(\mathbf{x}_0)$, it is the best local monomial approximation $g(\mathbf{x}_0)$ near x_0 , in the sense of first order Taylor approximation.

Algorithm: finds locally optimal power allocation

Initialize with feasible $\mathbf{z} = \{z_1, \dots, z_K\}$

1. Evaluate the denominator posynomial with the given \mathbf{z}
2. Compute for each term i in this posynomial

$$\alpha_i = \frac{\text{value of } i\text{th term in posynomial}}{\text{value of posynomial}}$$

3. Approximate the denominator of the posynomial ration by $\tilde{g}(\mathbf{z})$ using weights α_i
4. Solve the resulting GP
5. Go to step 1, using \mathbf{z} of step 4
6. Terminate the k^{th} loop if

$$||\mathbf{z}^{(k)} - \mathbf{z}^{(k-1)}|| \leq \epsilon$$

Outputs: A locally optimal \mathbf{z}

Subproblem 2 for HARQ-CC with slow fading

$$P_e \left(\sum_{\ell=1}^i E_\ell, \gamma \right) = \exp \left(\frac{-\sum_{\ell=1}^i E_\ell \cdot \gamma}{N_0} \right)$$
$$P_{\text{out}} = \mathbb{E}_\gamma \left\{ \prod_{\ell=1}^K P_e \left(\sum_{\ell=1}^i E_\ell, \gamma \right) \right\}$$

for Rayleigh fading

$$P_{\text{out}} = \frac{1}{\sum_{\ell=1}^i (K - \ell + 1) E_\ell}$$

Subproblem 2 for HARQ-CC with slow fading

$$\begin{aligned} & \max_{\{E_1, \dots, E_K\}} \quad \sum_{\ell=1}^K (K - \ell + 1) E_\ell \\ \text{s.t.} \quad & \sum_{\ell=1}^K E_\ell \left(\sum_{i=1}^{\ell-1} (\ell - i + 1) E_i + 1 \right)^{-1} \leq E_{av}^t \\ & \sum_{\ell=1}^K \mathbb{1}_{\{E_\ell \neq 0\}} E_\ell \left(\sum_{i=1}^{\ell-1} (\ell - i + 1) E_i + 1 \right)^{-1} \leq \frac{E_{av}^r}{R E_s} \\ & 0 \leq E_i \leq E_{\max} \end{aligned}$$

Converting to GP

Let $z_1 = 1 + E_1$, $z_2 = 1 + 2E_1 + E_2$, $z_3 = 1 + 3E_1 + 2E_2 + E_3$

$$E_i = z_i - 2z_{i-1} + z_{i-2}$$

$$\begin{aligned} & \max_{\{z_1, \dots, z_K\}} \quad z_K \\ \text{s.t.} \quad & \sum_{\ell=1}^K z_{\ell-1}^{-1} (z_\ell + z_{\ell-2}) \leq E_{av}^t - 2K + 1 \\ & \sum_{\ell=1}^K z_{\ell-1}^{-1} \leq \frac{E_{av}^r}{RE_s} \\ & 0 \leq z_i - 2z_{i-1} + z_{i-2} \leq E_{\max} \end{aligned}$$

Subproblem 1

- ▶ Can be solved using merit-based sequential quadratic programming (MSQP)
- ▶ MSQP guarantees global convergence to local optimum under some weak conditions.