

Throughput Capacities and Optimal Resource Allocation in Multiaccess Fading Channels

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- Multiaccess Fading Channel Model
- Capacity Under Power Control
- Polymatroid Structure and Greedy Algorithm
- Polymatroid Structure of Channel Capacity
- A Lagrangian Characterization of the Capacity Region
- Optimal Power and Rate Allocation for a Fading State
- Boundary of the Capacity Region
- Cases for throughput capacity and power allocation in fading channels

Preliminary:

The capacity region for classical discrete memoryless multiple access channel

$$(R_1, R_2) \quad \text{satisfying :} \quad R_1 < I(X_1; Y | X_2)$$
$$R_2 < I(X_2; Y | X_1)$$
$$R_1 + R_2 < I(X_1, X_2; Y)$$

with fixed probability transitions $p(y | x_1, x_2)$

and for some independent input distribution $p(x_1)p(x_2)$

In the case of the Gaussian multiple access channel $Y_i = X_{1i} + X_{2i} + Z_i$

$$R_1 \leq \frac{1}{2} \log\left(1 + \frac{P_1}{\sigma^2}\right)$$

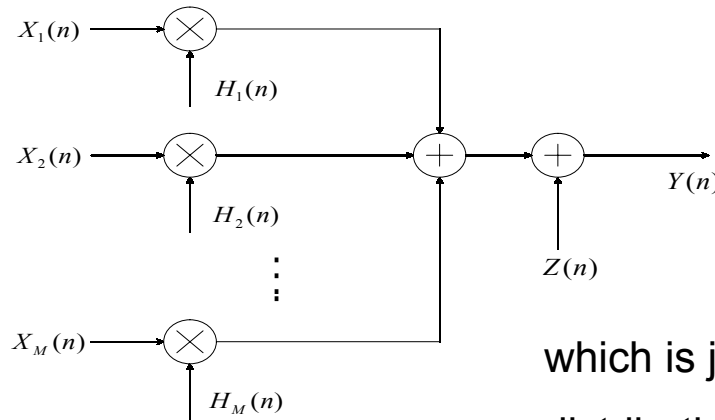
$$R_2 \leq \frac{1}{2} \log\left(1 + \frac{P_2}{\sigma^2}\right)$$

$$R_1 + R_2 \leq \frac{1}{2} \log\left(1 + \frac{P_1 + P_2}{\sigma^2}\right)$$

where $Z \sim N(0, \sigma^2)$, P_1, P_2 are the power constraint of X_1, X_2

The upper bounds are achieved when $X_1 \sim N(0, P_1)$, $X_2 \sim N(0, P_2)$

Multiaccess Fading Channel Model:



$$Y(n) = \sum_{i=1}^M \sqrt{H_i(n)} X_i(n) + Z(n)$$

$H_i(n)$ is the fading process of the i th user, which is jointly stationary and ergodic and whose stationary distribution has continuous density and is bounded

I. Capacity region of the fixed Gaussian multiaccess channel

$$C_g(\mathbf{h}, \mathbf{P}) = \left\{ \mathbf{R} : \mathbf{R}(S) \leq \frac{1}{2} \log \left(1 + \frac{\sum_{i \in S} h_i P_i}{\sigma^2} \right) \quad \text{for every } S \subset \{1, \dots, M\} \right\}$$

II. Capacity region of time-Varying channel with CSI on receiver side only

$$\left\{ (R_1, \dots, R_M) : \mathbf{R}(S) \leq \mathbf{E}_{\mathbf{H}} \left[\frac{1}{2} \log \left(1 + \frac{\sum_{i \in S} H_i P_i}{\sigma^2} \right) \right] \right\}, \quad \forall S \subset \{1, \dots, M\}$$

Capacity Under Power Control: (CSI on all transceivers)

Def: A power-control policy PC : a mapping of $\mathbf{h} = (h_1, \dots, h_M) \rightarrow R_+^M$

I. Capacity region under a power-control policy

$$C_f(PC) \equiv \{\mathbf{R} : \mathbf{R}(S) \leq \mathbf{E}_H[\frac{1}{2} \log(1 + \frac{1}{\sigma^2} \sum_{i \in S} H_i PC_i(\mathbf{H}))]\}, \quad \forall S \subset \{1, \dots, M\}$$

II. Throughput capacity region under power control

Theorem 2.1: $C(\bar{\mathbf{P}}) \equiv \bigcup_{PC \in F} C_f(PC)$

where $F \equiv \{PC : \mathbf{E}_H[PC(\mathbf{H})] \leq \bar{\mathbf{P}} \quad \forall i\}$

• achievability: for any PC , look the channel as unit power channel with $h_i PC_i(\mathbf{h})$,

$C_f(PC)$ is achievable and $\bigcup_{PC \in F} C_f(PC) \subset C(\bar{\mathbf{P}})$

• converse: By Fano's inequality.

Polymatroid Structure and Greedy Algorithm

I. Polymatroid Structure

Def: $E = \{1, \dots, M\}$ and $f : 2^E \rightarrow R_+$ be a set function. The polyhedron $B(f) \equiv \{(x_1, \dots, x_M) : \mathbf{x}(S) \leq f(S) \forall S \subset E, x_i \geq 0 \forall i\}$ is a polymatroid if the set function

- f satisfies
- 1) $f(\emptyset) = 0$ (normalized).
 - 2) $f(S) \leq f(T)$ if $S \subset T$ (nondecreasing).
 - 3) $f(S) + f(T) \geq f(S \cup T) + f(S \cap T)$ (submodular)

II. Greedy Algorithm

optimization problem $\max \mathbf{1} \cdot \mathbf{x}$ subject to $\mathbf{x} \in B(f)$

- **Initialization:** Set $x_i = 0$ for all i . Set $k = 1$
- **Step k :** Increase the value of $x_{\pi^*(k)}$ until a constraint becomes tight. Goto Step $k + 1$
- After M steps, the optimal solution is reached.

Polymatroid Structure of Channel Capacity

I. Classical discrete memoryless multiaccess channel

$\{\mathbf{R} \in R_+^M : \mathbf{R}(S) \leq I[Y; \mathbf{X}(S) | \mathbf{X}(S^c)] \forall S \subset E\}$ **is a polymatroid.**

II. Memoryless Gaussian multiaccess channel

$C_g(\mathbf{h}, \mathbf{P})$ **is a polymatroid.**

III. A power control policy PC , $C_f(PC)$ **is a polymatroid**

Def: A rate allocation policy R is a mapping: $\mathbf{h} \rightarrow R_+^M$

for each fading state \mathbf{h} , $R_i(\mathbf{h})$ can be interpreted as the rate allocated to user i

IV. For any power control policy PC

$C_f(PC) = \{\mathbf{E}_{\mathbf{H}}[R(\mathbf{H})] : R \text{ is a rate allocation policy s.t. } \forall \mathbf{h} R(\mathbf{h}) \in C_g(\mathbf{h}, PC(\mathbf{h}))\}$

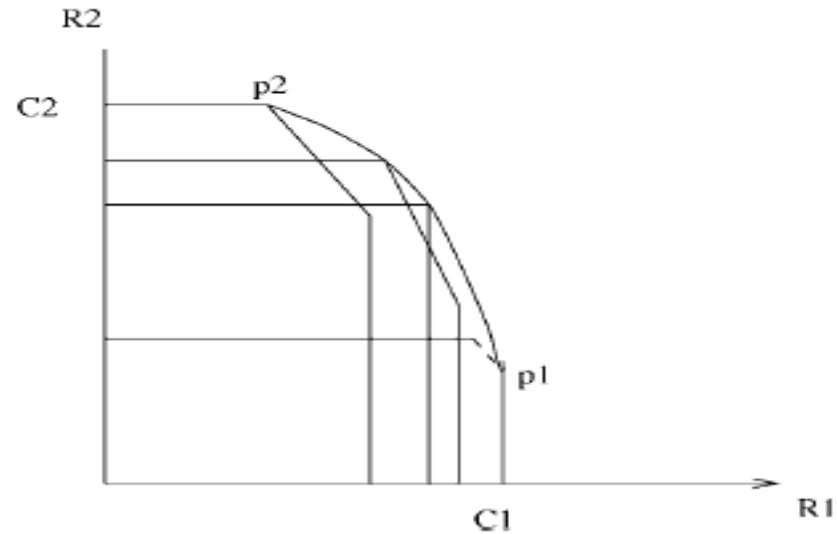
Furthermore, $\mathbf{v}(\pi) = \mathbf{E}_{\mathbf{H}}[\mathbf{v}_{\mathbf{H}}(\pi)]$, for any permutation π on E ,

Where $\mathbf{v}(\pi) \leftrightarrow C_f(PC)$,

and $\mathbf{v}_{\mathbf{h}}(\pi) \leftrightarrow C_g(\mathbf{h}, PC(\mathbf{h}))$,

for each state \mathbf{h} , corresponding to same permutation π

A Lagrangian Characterization of the Capacity Region



Capacity Region: $C_f(PC)$ and $C(\bar{\mathbf{P}})$

Def: The *boundary surface* of $C(\bar{\mathbf{P}})$: is the set of those rates such that no component can be increased with the other components remaining fixed, while remaining in $C(\bar{\mathbf{P}})$

Lemma 3.10: I. The boundary surface of $C(\bar{\mathbf{P}})$ is the closure of all points \mathbf{R}^* such that \mathbf{R}^* is a solution to the optimization problem:

$$\max_{\mathbf{R}} \mathbf{u} \cdot \mathbf{R} \quad \text{subject to} \quad \mathbf{R} \in C(\bar{\mathbf{P}}) \quad \text{for some} \quad \mathbf{u} \in R_+^M$$

II. For a given \mathbf{u} , \mathbf{R}^* is a solution to the above problem if and only if there exists a $\mathbf{l} \in R_+^M$, rate allocation policy $R(\cdot)$ and power control policy $PC(\cdot)$ such that for every joint fading-state \mathbf{h} , $(R(\mathbf{h}), PC(\mathbf{h}))$ is a solution to the optimization problem:

$$\max_{(\mathbf{r}, \mathbf{p})} \mathbf{u} \cdot \mathbf{r} - \mathbf{l} \cdot \mathbf{p} \quad \text{subject to} \quad \mathbf{r} \in C_g(\mathbf{h}, \mathbf{p})$$

and $\mathbf{E}_{\mathbf{H}}[R_i(\mathbf{H})] = R_i^*$, $\mathbf{E}_{\mathbf{H}}[PC_i(\mathbf{H})] = \bar{P}_i$, $i = 1, \dots, M$

where \bar{P}_i is the constraint on the average power of user i

\mathbf{u}	-----	rate rewards
\mathbf{l}	-----	power prices
\mathbf{h}	-----	joint fading state
$\bar{\mathbf{P}}$	-----	average power constraints
$(R(\mathbf{h}), PC(\mathbf{h}))$		

Optimal Power and Rate Allocation for a Fading State h

Theorem 3.14: generic problem

$$\max_{(\mathbf{x}, \mathbf{y})} \mathbf{u} \cdot \mathbf{x} - \mathbf{l} \cdot \mathbf{y} \quad \text{subject to } \mathbf{x}(S) \leq g(\mathbf{y}(S))$$

g : monotonically increasing concave function

Define the marginal utility functions

$$u_i(z) \equiv u_i g'(z) - l_i, \quad i = 1, \dots, M$$

$$u^*(z) \equiv [\max_i u_i(z)]^+$$

The solution to the problem is:

$$\int_0^\infty u^*(z) dz$$

Time-invariant Gaussian channel

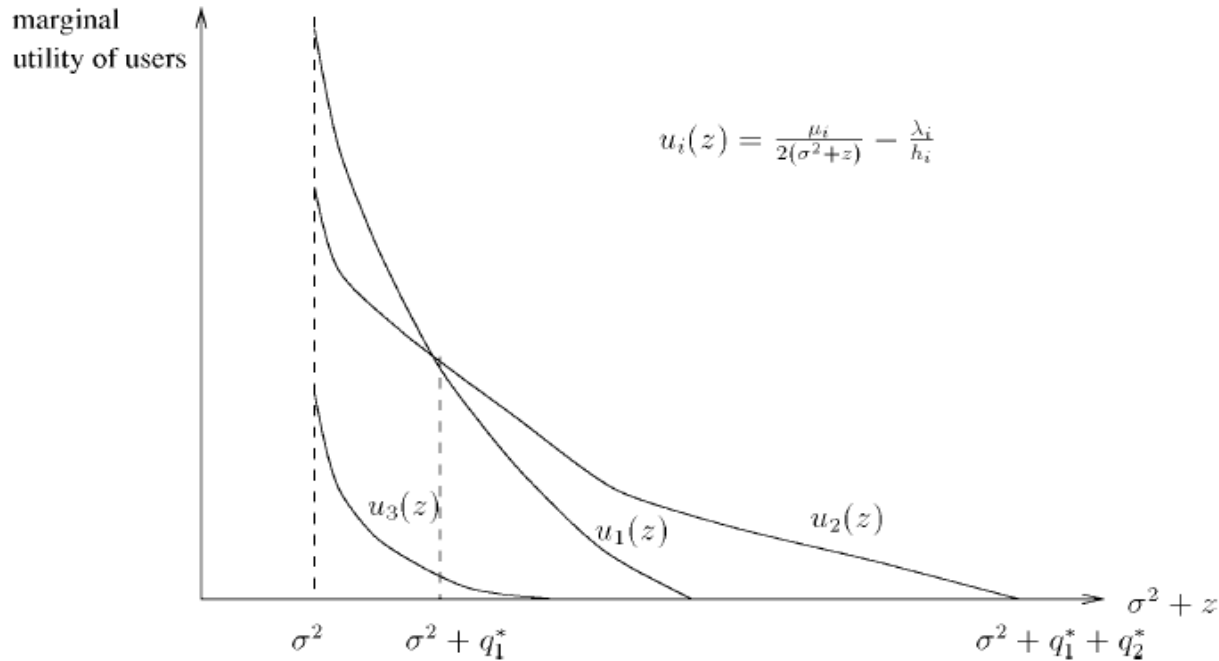
$$\max \sum_i u_i r_i - \sum_i \frac{l_i}{h_i} q_i \quad \text{subject to } \mathbf{r}(S) \leq g(\mathbf{q}(S))$$

$$g(z) \equiv \frac{1}{2} \log\left(1 + \frac{z}{\sigma^2}\right)$$

$$u_i(z) \equiv \frac{u_i}{2(\sigma^2 + z)} - \frac{l_i}{h_i}$$

$$u^*(z) \equiv [\max_i u_i(z)]^+$$

$$\int_0^\infty u^*(z) dz$$



A three-user example illustrating the greedy power allocation

With probability 1, the optimal power and rate allocation is unique and is explicitly given by

$$R_i^*(\mathbf{h}) = \int_{A_i} \frac{1}{2(\sigma^2 + z)} dz \quad PC_i^*(\mathbf{h}) = |A_i|$$

where

$$A_i \equiv \{z \in [0, \infty) : u_i(z) > u_j(z) \quad \forall j \neq i \text{ and } u_i(z) > 0\}$$

Boundary of the Capacity Region $C(\bar{\mathbf{P}})$

I. *Lemma 3.15:* (Uniqueness) for $\mathbf{u} \in R_+^M$, there is a unique \mathbf{R}^* on the boundary which maximizes $\mathbf{u} \cdot \mathbf{R}$, and there is a unique Lagrangian power price $\mathbf{1}$ such that the optimal power allocation satisfies the average power constraints

II. *Theorem 3.16:* For independent users' fading processes, the boundary of $C(\bar{\mathbf{P}})$ is the closure of the parametrically defined surface

$$\{\mathbf{R}^*(\mathbf{u}) : \mathbf{u} \in R_+^M, \sum_i u_i = 1\}$$

where

$$R_i^*(\mathbf{u}) = \int_0^\infty \frac{1}{2(\sigma^2 + z)} \left\{ \int_{\frac{2l_i(\sigma^2+z)}{u_i}}^\infty \prod_{k \neq i} F_k \left(\frac{2l_k h(\sigma^2 + z)}{2l_i(\sigma^2 + z) + (u_k - u_i)h} \right) f_i(h) dh \right\} dz$$

and where the vector $\mathbf{1}$ is the unique solution of the equations

$$\int_0^\infty \left\{ \int_{\frac{2l_i(\sigma^2+z)}{u_i}}^\infty \frac{1}{h} \times \prod_{k \neq i} F_k \left(\frac{2l_k h(\sigma^2 + z)}{2l_i(\sigma^2 + z) + (u_k - u_i)h} \right) f_i(h) dh \right\} dz = \bar{P}_i$$

F_i and f_i are the cdf and pdf of the stationary fading distribution of user i

Cases for throughput capacity and power allocation in fading channels

I. Single-User Channel

$$R^* = \int_0^\infty \frac{1}{2(\sigma^2 + z)} \left\{ \int_{\frac{2\lambda(\sigma^2+z)}{u}}^\infty f(h) dh \right\} dz = \int_0^\infty \frac{1}{2} \log\left(1 + \frac{h}{\sigma^2} \left(\frac{u}{2l} - \frac{\sigma^2}{h}\right)^+\right) f(h) dh$$

$$\frac{u}{2l} \text{ satisfies the power constraint } \int_0^\infty \left(\frac{u}{2l} - \frac{\sigma^2}{h}\right)^+ f(h) dh = \bar{P}$$

Note:

- Time Water-Filling Solution: Power allocation over a set of parallel single-user channels, one for each fading level h
- More power is used when the channel is good and little or even no power when it is bad.

II. Maximum Sum-Rate Point $(u_1 = \dots = u_M = 1)$

Utility Functions:
$$u_i(z) = \frac{1}{2(\sigma^2 + z)} - \frac{l_i}{h_i}$$

Power Control Strategy:
$$PC_i^*(\mathbf{h}, \mathbf{l}) = \begin{cases} \left(\frac{1}{2l_i} - \frac{\sigma^2}{h_i}\right)^+, & \text{if } h_i > \frac{l_i}{l_j} h_j \text{ for all } j \\ 0, & \text{else.} \end{cases}$$

Optimal Rates :
$$R_i^* = \int_0^\infty \frac{1}{2} \log\left(1 + \frac{h}{\sigma^2} \left(\frac{1}{2\lambda_i} - \frac{\sigma^2}{h}\right)^+\right) \times \prod_{k \neq i} F_k\left(\frac{l_k h}{l_i}\right) f(h) dh$$

Power Prices Constant:
$$\int_0^\infty \left(\frac{1}{2l_i} - \frac{\sigma^2}{h}\right)^+ \prod_{k \neq i} F_k\left(\frac{l_k h}{l_i}\right) f(h) dh = \bar{P}_i,$$

Note:

- Here the optimal rates are in the sense of sum of all users' rates.
- The optimal (TDMA) strategy allows at most one user to transmit at any given fading state, this lucky user has the best channel and largest available power.
- As for the two-user example, the boundary point corresponding to $u_1 = u_2$ is the corner point of a rectangular $C_f(PC)$

III. *Multiple Classes of Users*

If the fading processes of the users have very different statistics, in order to equalize users' rates, unequal rate awards can be assigned to users.

Two classes users example:

Class 1: *far* users at the cell boundary, all assigned rate award u_1

Class 2: *near* users close to the base station, all assigned rate award u_2

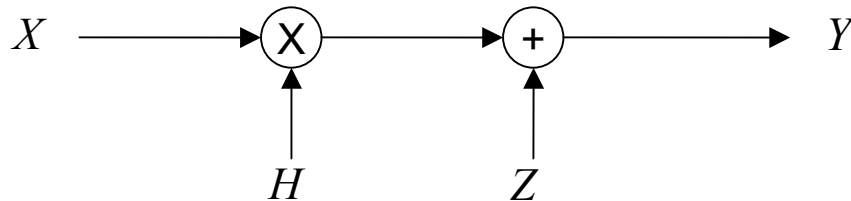
$$u_1 > u_2$$

Note:

- At each fading state, only the strongest user in each class transmits.
- The two strongest users are decoded by successive cancellation, with nearby user decoded first.

Appendix:

The channel capacity for the receiver tracking the channel only.



$$C = \max_{p(x)} I(X; Y, H) = \max_{p(x)} [I(X; H) + I(X; Y | H)]$$

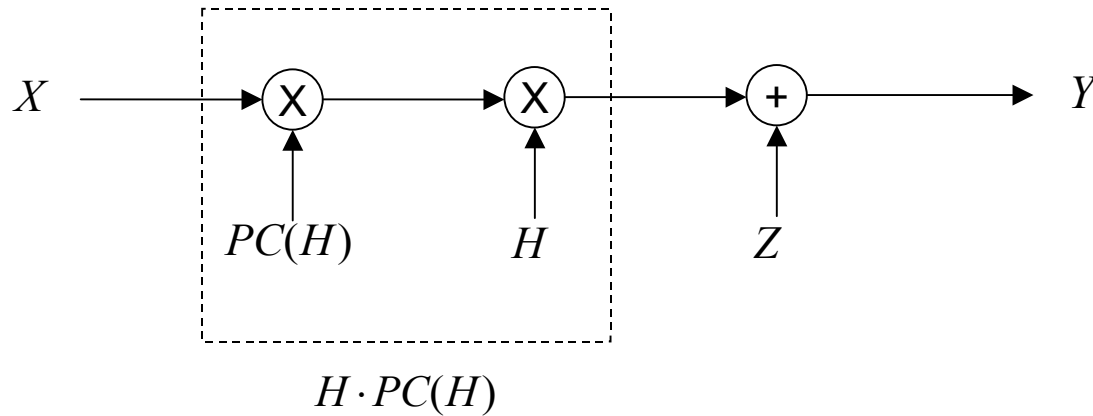
($I(X; H) = 0$ --- the fading variable and the channel input are independent)

$$= \max_{p(x)} E[I(X; Y | H = h)]$$

$$\{(R_1, \dots, R_M) : \mathbf{R}(S) \leq \mathbf{E}_{\mathbf{H}} \left[\frac{1}{2} \log \left(1 + \frac{\sum_{i \in S} H_i P_i}{\sigma^2} \right) \right], \quad \forall S \subset \{1, \dots, M\}\}$$

Appendix:

Capacity region under a power-control policy



$$C_f(PC) \equiv \{\mathbf{R} : \mathbf{R}(S) \leq \mathbf{E}_H \left[\frac{1}{2} \log \left(1 + \frac{1}{\sigma^2} \sum_{i \in S} H_i PC_i(\mathbf{H}) \right) \right], \quad \forall S \subset \{1, \dots, M\}\}$$

Appendix:

$B(f) \equiv \{(x_1, \dots, x_M) : \mathbf{x}(S) \leq f(S) \forall S \subset E, x_i \geq 0 \forall i\}$ is a polymatroid

π is a permutation on the set E , define the vector $\mathbf{v}(\pi) \in R^M$ by

$$v_{\pi(1)}(\pi) = f(\pi(1))$$

$$v_{\pi(i)}(\pi) = f(\{\pi(1), \dots, \pi(i)\}) - f(\{\pi(1), \dots, \pi(i-1)\}) \quad i = 2, \dots, M.$$

Then $\mathbf{v}(\pi)$ is a vertex of $B(f)$ for every permutation π

Conversely, suppose f is a set function and $B(f)$ is the polyhedron defined as above.

Then if $\mathbf{v}(\pi) \in B(f)$ for every permutation π , then $B(f)$ is a polymatroid.

Consider the polyhedron $\{\mathbf{R} \in R_+^M : \mathbf{R}(S) \leq I[Y; \mathbf{X}(S) | \mathbf{X}(S^c)] \forall S \subset E\}$

$$f(S) = I[Y; \mathbf{X}(S) | \mathbf{X}(S^c)]$$

π is a permutation on the set E , define the vector $\mathbf{R}(\pi) \in R^M$ by

$$R_{\pi(1)}(\pi) = I[Y; X_{\pi(1)} | \mathbf{X}(\{\pi(2), \dots, \pi(M)\})]$$

$$R_{\pi(i)}(\pi) = I[Y; X_{\pi(i)} | \mathbf{X}(\{\pi(i+1), \dots, \pi(M)\})], \quad i = 1, \dots, M-1 \quad (\text{Chain rule})$$

$$R_{\pi(M)}(\pi) = I[Y; X_{\pi(M)}]$$

Obviously, $\mathbf{R}(\pi)$ lies in $\mathbf{R}(S)$, so the polyhedron region is a polymatroid

Appendix:

A Lagrangian multiplier:

Optimization problem: $\max_{\mathbf{R}} \mathbf{u} \cdot \mathbf{R}$ subject to $\mathbf{R} \in C(\bar{\mathbf{P}})$

\Downarrow

$\max_{\mathbf{R} \in C(\mathbf{P})} \mathbf{u} \cdot \mathbf{R}$ subject to $\mathbf{P} = \bar{\mathbf{P}}$

Lagrangian multipliers

\Downarrow

there exists an \mathbf{l} , $\max_{\mathbf{R} \in C(\mathbf{P})} \mathbf{u} \cdot \mathbf{R} - \mathbf{l} \cdot (\mathbf{P} - \bar{\mathbf{P}})$

\Downarrow

there exists an \mathbf{l} , $\max_{\mathbf{R} \in C(\mathbf{P})} \mathbf{u} \cdot \mathbf{R} - \mathbf{l} \cdot \mathbf{P}$

Appendix:

The classic water-filling solution for the single-user case

For each fading state h , the optimization problem:

$$\max_{r,q} [r - \frac{l}{h}q] \quad \text{subject to} \quad r \leq \frac{1}{2} \log(1 + \frac{q}{\sigma^2})$$

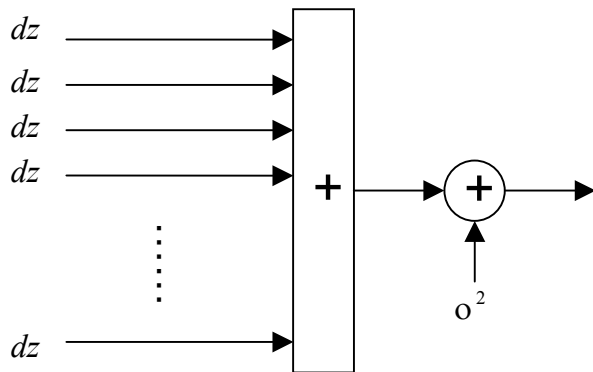
$$\Downarrow$$

$$\max_q [\frac{1}{2} \log(1 + \frac{q}{\sigma^2}) - \frac{l}{h}q]$$

$$\Downarrow \quad \longleftarrow \quad \frac{1}{2} \log(1 + \frac{q}{\sigma^2}) = \int_0^q \frac{1}{2(\sigma^2 + z)} dz$$

$$\max_q \int_0^q [\frac{1}{2(\sigma^2 + z)} - \frac{l}{h}] dz$$

received power



marginal utility function $u(z) \equiv \frac{1}{2(\sigma^2 + z)} - \frac{l}{h}$

Optimization problem turns out to be :

Adding more virtual users until $u(z) \cdot dz < 0$