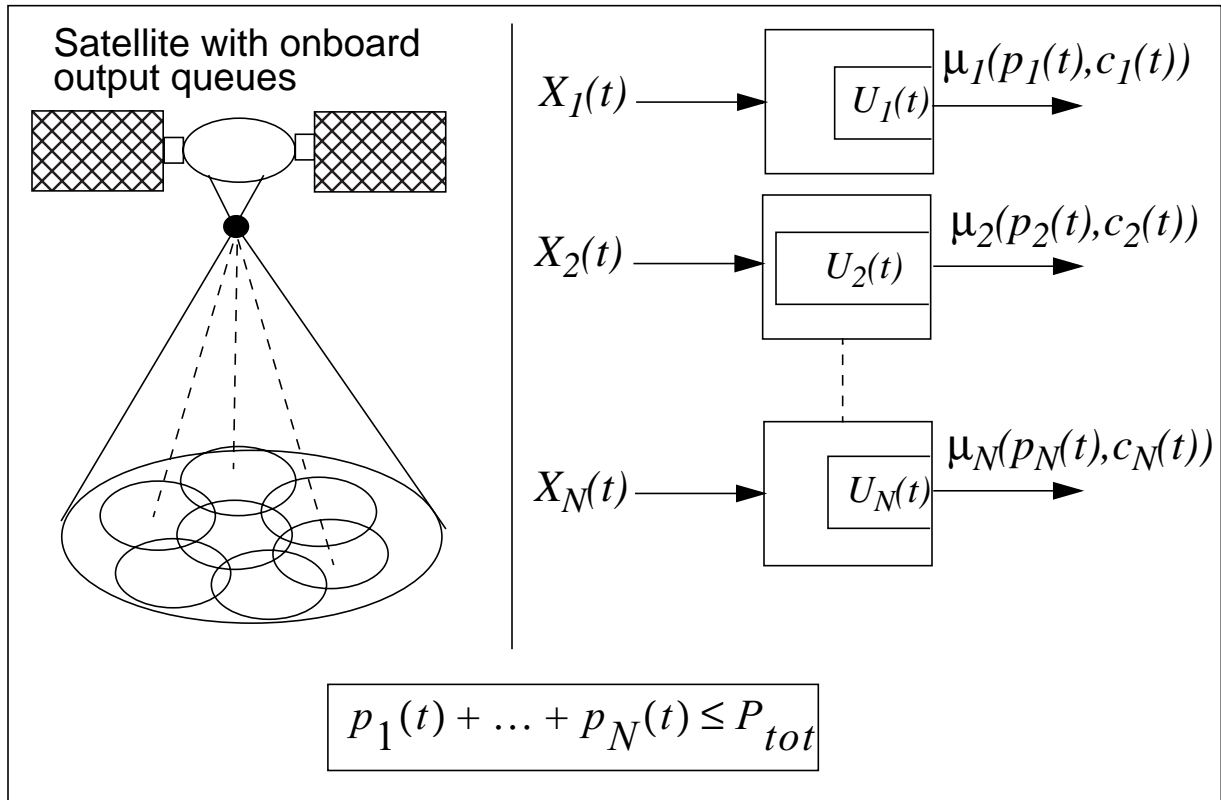


# Power and Server Allocation in a Multi-Beam Satellite with Time Varying Channels

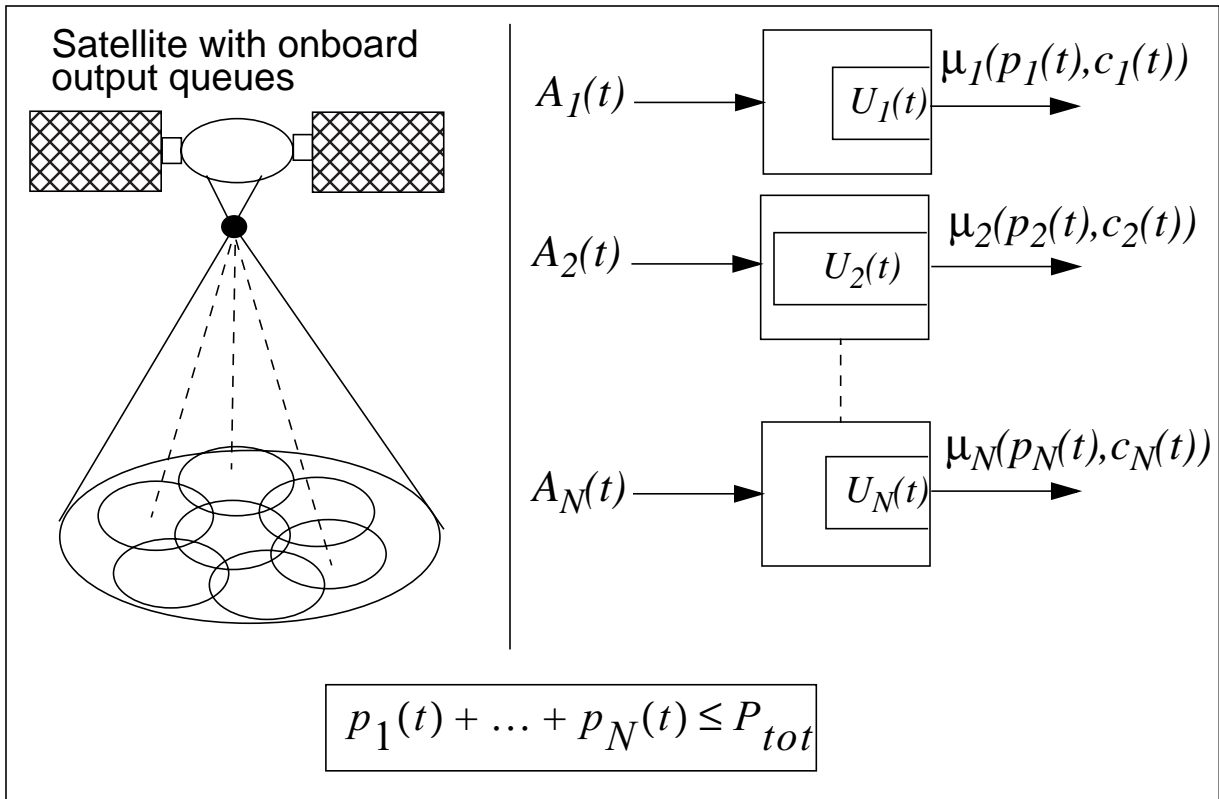


Michael Neely  
MIT -- LIDS  
mjneely@mit.edu

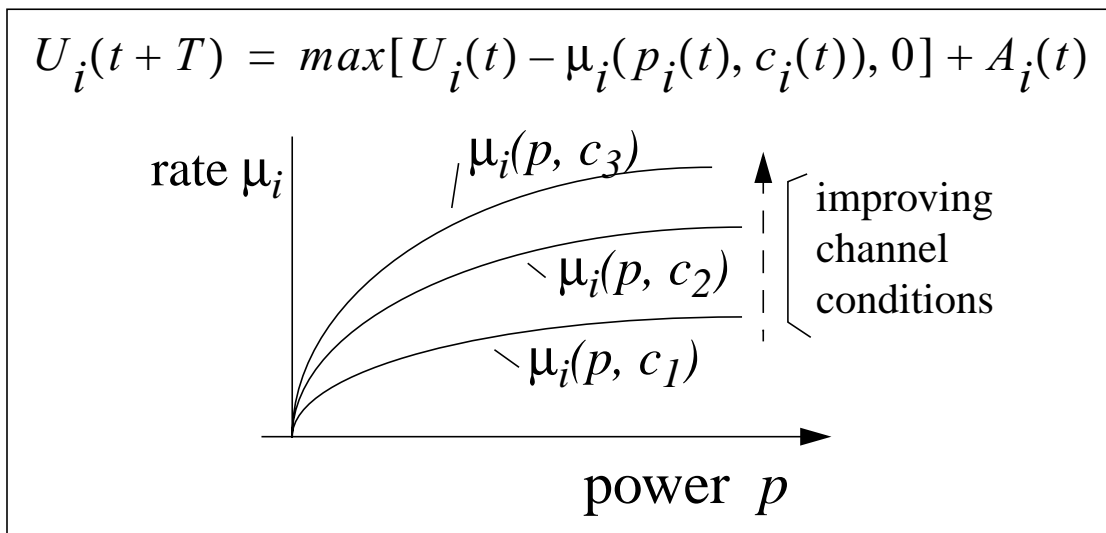
Eytan Modiano -- LIDS  
Charlie Rohrs -- LIDS

Conference: INFOCOM 2002

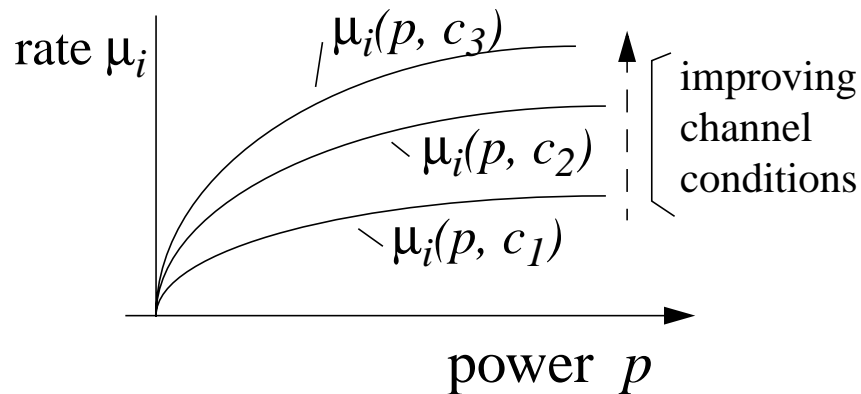
Journal: M.J. Neely, E. Modiano, and C. E. Rohrs, "Power Allocation and Routing in Multi-Beam Satellites with Time Varying Channels," IEEE Transactions on Networking, Feb. 2003.



1. Data Arrives as Random Processes  $\{A_i(t)\}$  (rates  $\{\lambda_i\}$ )
2. Time Varying Channels  $\vec{C}(t) = (c_1(t), c_2(t), \dots, c_N(t))$
3. Rate-Power curves  $\mu_i(p_i, c_i)$



## Link Budget Curves: (Concave, Continuous)

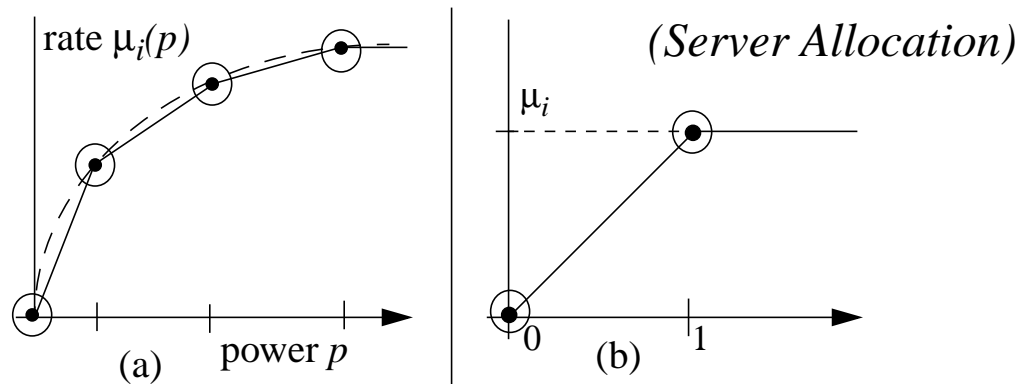


### Curve Examples:

>  $\log(1 + p_i c_i)$  ( $c_i \sim$  attenuation-to-noise coefficient)

> Any curves for codes designed to achieve a specified low probability of error  $\delta$ .

Curves may have a finite set of feasible rate-power points (corresponding to a finite databank of codes).



Create a *Virtual Power Curve*  $\tilde{\mu}_i(p_i, c_i) =$  Piecewise Linear Interpolation of Feasible Points.

Goals:

-Develop Capacity region  $\Omega = \{(\lambda_1, \lambda_2, \dots, \lambda_N) \text{ rates that the system can stably support}\}$ .

(We consider general ergodic arrival processes, and all possible power allocation strategies)

-Develop a Dynamic Power Allocation Policy to stabilize system and thereby achieve maximum throughput and maintain acceptably low levels of unfinished work backlog in all queues.

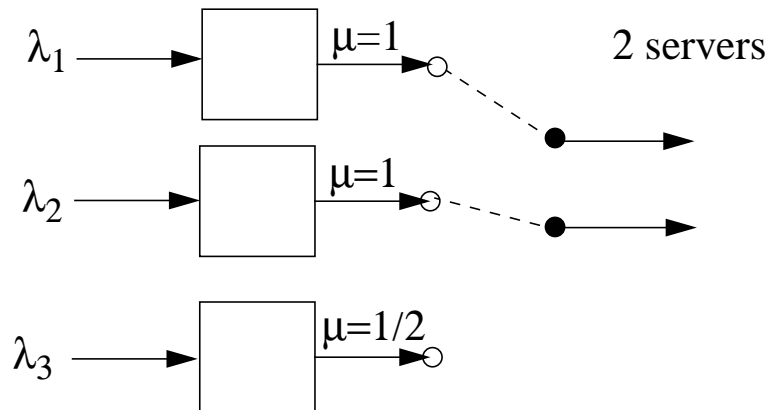
(Consider Timeslotted structure, *iid* assumptions  
Can be generalized to Markovian inputs/Channel states)

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Example: Special Case of *Server Allocation*:

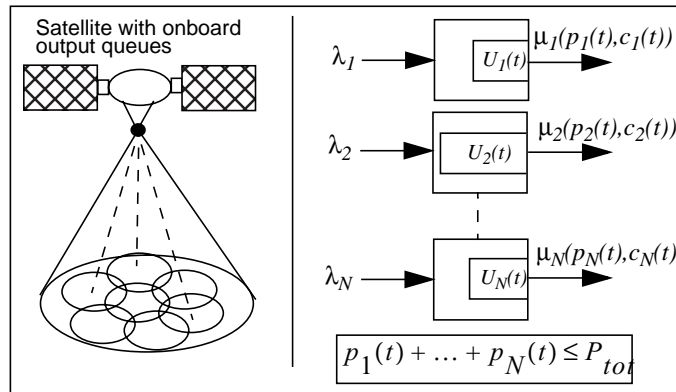
2 servers, 3 queues (static channel conditions)

Packet arrives to queue  $i$  every timeslot with probability  $\lambda_i$



Serving the 2 fastest, non-empty queues does not stabilize the system in this case...

$$(\lambda_1 = \lambda_2 = p, \quad \lambda_3 = \frac{1}{2}(1 - p^2) + \epsilon)$$



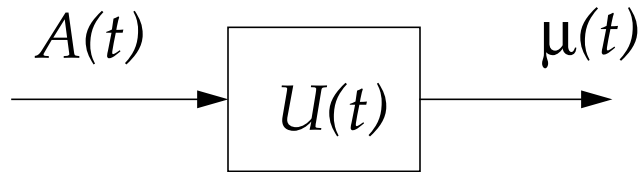
## Definition of Capacity Region $\Omega$ :

Let  $\lambda_i$  be the bit rate of stream  $X_i(t)$

The Capacity Region  $\Omega$  is the set of all rate vectors  $\vec{\lambda} = (\lambda_1, \dots, \lambda_N)$  such that:

- The network is necessarily unstable whenever  $\vec{\lambda} \notin \Omega$ .
- The network can be stabilized if  $\vec{\lambda}$  is *strictly interior* to  $\Omega$ .

## A note on Stability:



$A(t)$  = Arrival Process (assumed ergodic of rate  $\lambda$ ).

$\mu(t)$  = instantaneous processing rate (potentially non-ergodic).

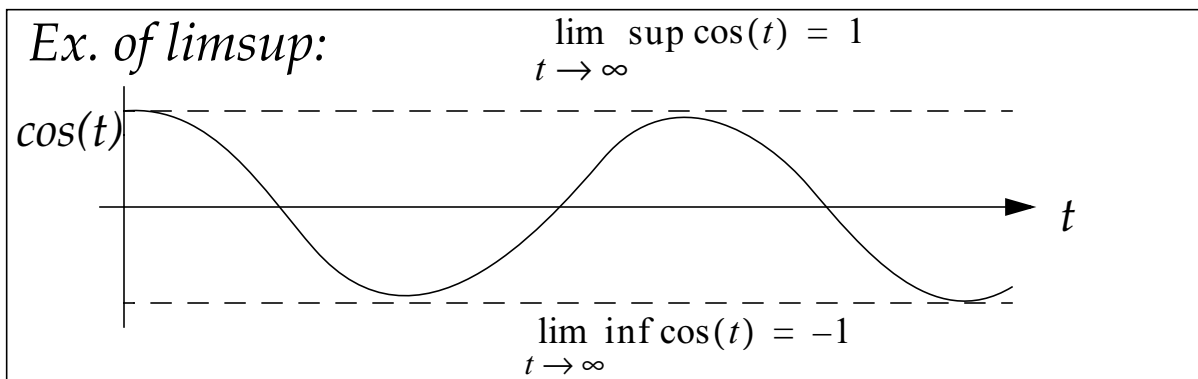
$U(t)$  = Unfinished work (bits) in queue at time  $t$ .

*Definition:* The *overflow function*  $g(M)$ :

$$g(M) = \limsup_{t \rightarrow \infty} \left[ \frac{1}{t} \int_0^t 1_{\{U(\tau) > M\}} d\tau \right]$$

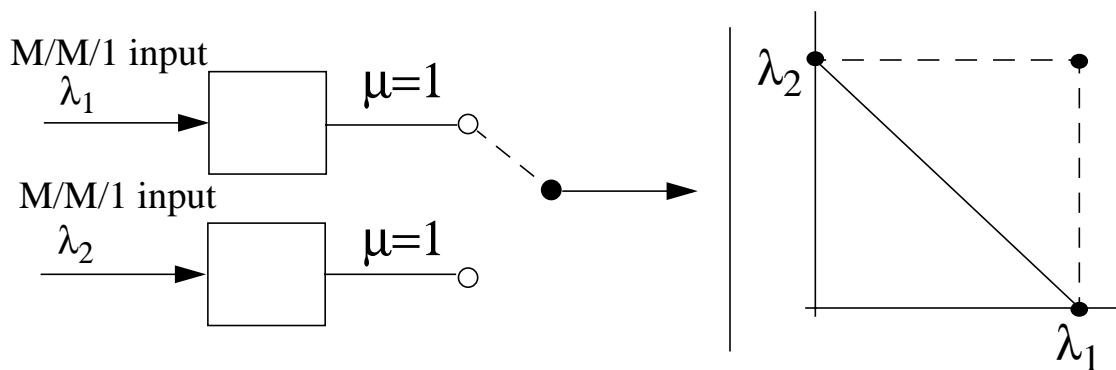
$g(M)$  represents the average fraction of time the unfinished work is above the level  $M$ .

Definition: A queueing system is stable if  $g(M) \rightarrow 0$  as  $M \rightarrow \infty$ .



What can go wrong with wrong def.?

A network is stable if all queues are stable

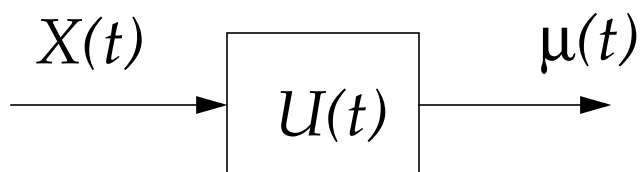


The *lim sup* definition is essential to obtain the correct notion of stability.

The above system is stable whenever  $\lambda_1 + \lambda_2 < 1$ .

If *lim inf* is used, it is stable for  $\lambda_1 + \lambda_2 < 2$ .

Necessary Condition (Single Queue):

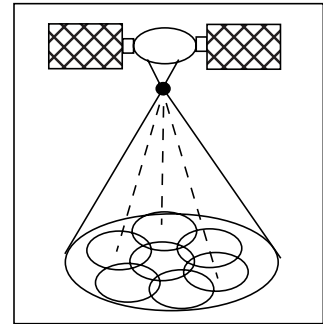


Define:  $\underline{\mu} = \lim_{t \rightarrow \infty} \inf \frac{1}{t} \int_0^t \mu(\tau) d\tau$

Lemma--Necessary Cond. for Stability:

$$\lambda \leq \underline{\mu} \quad \square$$

Derivation of Capacity Region  $\Omega$   
of Satellite Downlink: The  
Necessary Condition  
 (consider constant channel case)



Suppose the inputs have rates  $(\lambda_1, \dots, \lambda_N)$ , and some power allocations  $\{p_i(t)\}$  stabilize the system (perhaps designed with full knowledge of future).

$\{p_i(t)\}$  satisfy instantaneous power constraint:

$$\sum_{i=1}^N p_i(t) \leq P_{tot} \quad \forall t$$

Define:

$$\underline{\mu}_i = \limsup_{t \rightarrow \infty} \frac{1}{t} \int_0^t \mu_i(p_i(\tau)) d\tau$$

By stability:

$$\begin{aligned} \lambda_i \leq \underline{\mu}_i &\leq \frac{1}{t^*} \int_0^{t^*} \mu_i(p_i(\tau)) d\tau + \varepsilon \\ &\leq \mu_i \left( \frac{1}{t^*} \int_0^{t^*} p_i(\tau) d\tau \right) \quad (\text{by concavity}) \\ &= \mu(p_i^*) \quad (\text{and } p_i^* \text{ satisfies power constraints}) \quad \square \end{aligned}$$



Similarly, for time varying channels:

We can restrict ourselves to stationary power allocation policies (allocate fixed power vector

$(P_1^{\vec{C}}(t), \dots, P_N^{\vec{C}}(t))$  whenever in state  $\vec{C}(t)$ ).

Let:  $\pi_{\vec{C}}$  = Steady state probability of being in channel state  $\vec{C}$ .

The capacity region  $\Omega$  of the satellite downlink is the set of all rates  $(\lambda_1, \dots, \lambda_N)$  such that there exist power values  $p_i^{\vec{C}}$ :

$$\sum_{i=1}^N p_i^{\vec{C}}(t) \leq P_{tot} \quad (\text{Power Constraint})$$

$$\lambda_i \leq \sum_{\vec{C}} \pi_{\vec{C}} \mu_i(p_i^{\vec{C}}, c_i) \quad (\text{Rate inequality})$$

Would like to stabilize system without knowing channel statistics, input processes, or input rates  $(\lambda_1, \dots, \lambda_N)$ .

A Stabilizing Policy that considers  $\vec{C}(t)$  and  $\vec{U}(t)$ :

(consider timeslotted system under independence assumptions)

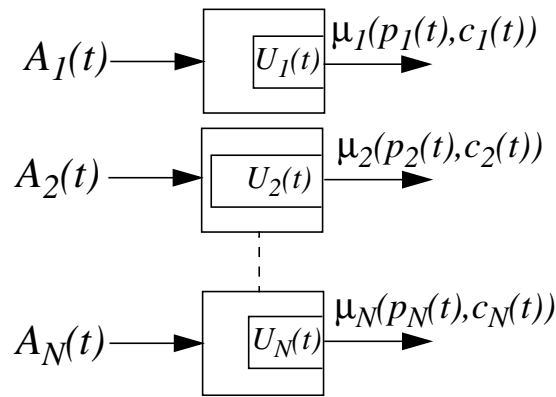
-Packets arrive to queue  $i$  with bit rate  $\lambda_i$

(Bit arrivals  $a_i \sim f_i(a_i)$  iid every timeslot,  $E[a_i] = \lambda_i$ ).

-Channel states  $\vec{C}(t)$  change every timeslot

(iid with probabilities  $\pi_{\vec{C}}$ )

- $U_i(t)$  = Unfinished work in node  $i$  at time  $t$ .



Strategy: Every timeslot, observe  $\vec{U}(t)$  and  $\vec{C}(t)$ :

Allocate  $\{p_i\}$  to Maximize: 
$$\sum_{i=1}^N \theta_i U_i(t) \mu_i(p_i, c_i(t))$$

Subject to: 
$$\sum_{i=1}^N p_i \leq P_{tot}$$

(the  $\theta_i$ 's are arbitrary positive weights for priority service)

The policy is stabilizing whenever the input rate vector  $\vec{\lambda}$  is strictly interior to the capacity region  $\Omega$ .

Analysis technique uses Lyapunov Drift (Lyapunov techniques well known in switching/scheduling literature [McKeown, Tassiulas, Leonardi]).

Dynamic Equation:

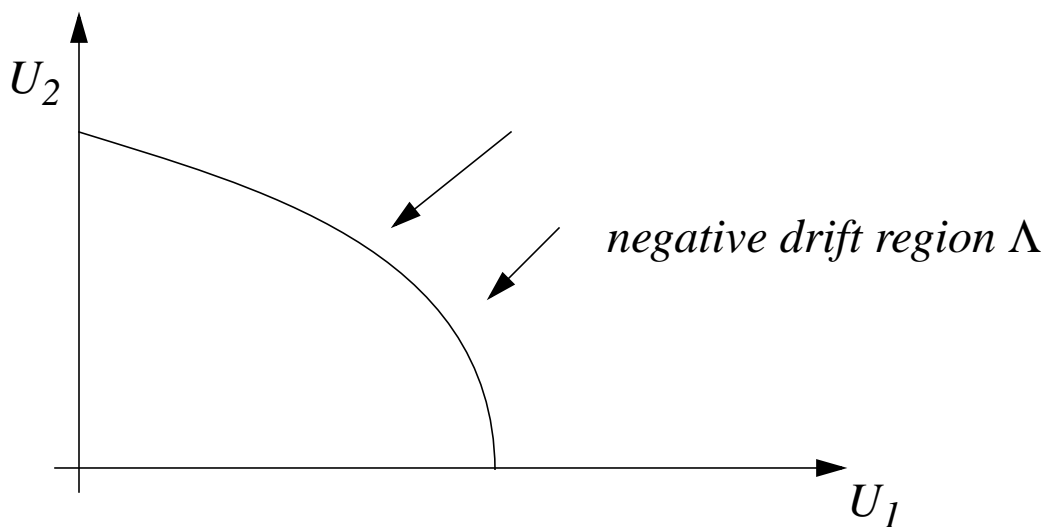
$$U_i(t + T) = \max[U_i(t) - \mu_i(p_i(t), c_i(t)), 0] + A_i(t)$$

Define Lyapunov Function:

$$L(\vec{U}) = \sum_{i=1}^N U_i^2$$

Can show drift in Lyapunov function is negative whenever  $\vec{U}(t)$  is outside some bounded region of the unfinished work state space:

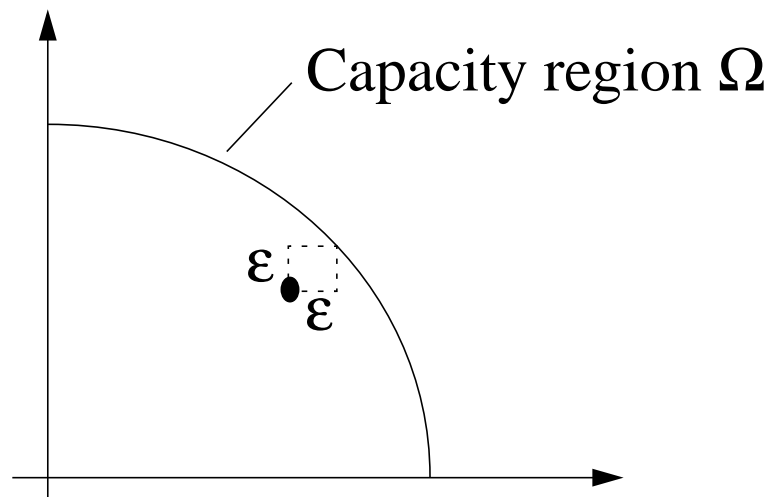
$$E[L(\vec{U}(t + T)) - L(\vec{U}(t)) | \vec{U}(t) \in \Lambda] \leq -\delta$$



### Delay Bound:

Suppose the rate vector  $\vec{\lambda}$  is strictly interior to  $\Omega$  so that a positive value  $\varepsilon$  can be added to each entry such that:

$$(\lambda_1 + \varepsilon, \dots, \lambda_N + \varepsilon) \in \Omega$$



$\varepsilon$  can be viewed as the “distance” to the boundary of the capacity region  $\Omega$ .

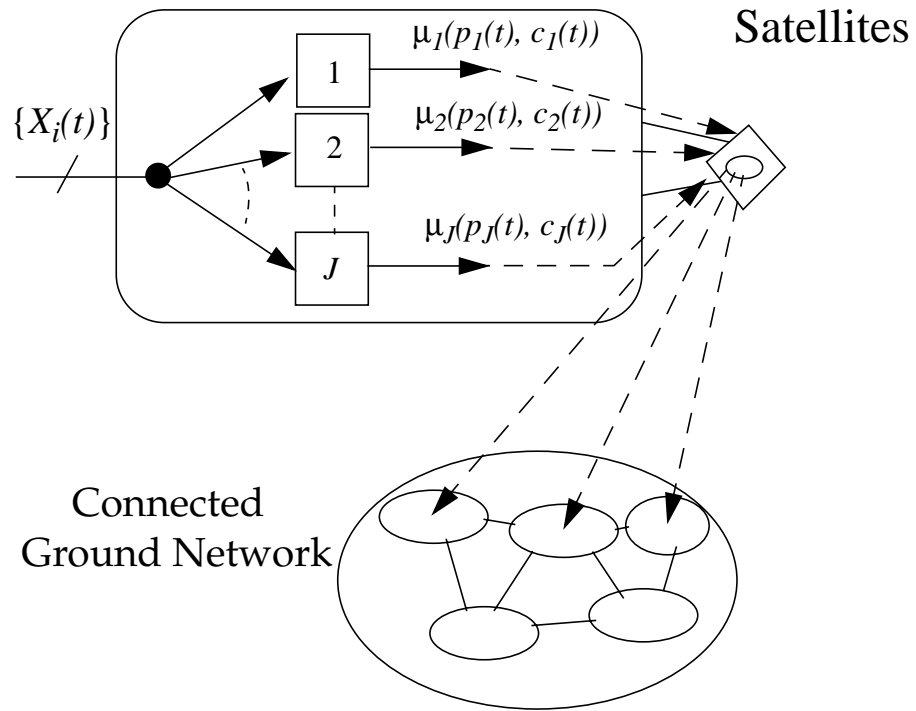
$$\text{Average Delay} \leq \frac{T \sum_i (E[A_i^2] + E[\mu_i^2]) / \lambda_{tot}}{2\varepsilon}$$

Note Fundamental Similarity to M/G/1 queue:

$$\text{Average Delay}_{(M/G/1)} = \frac{\lambda E(L^2) / \mu}{2(\mu - \lambda)}$$

$2^{nd}$  moment of packet length

## Joint Routing and Power Allocation:



Capacity Region:

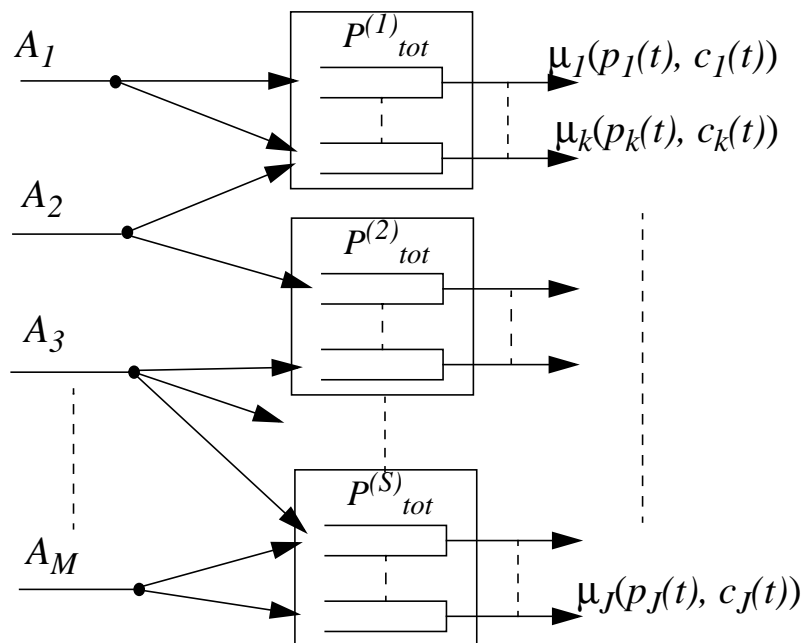
$$\lambda_1 + \dots + \lambda_J \leq \bar{\mu}_{out}$$

$$\bar{\mu}_{out} = \sum_{\vec{C}} \pi_{\vec{C}} \sum_{s=1}^S \max_{\left\{ \sum_{j \in Sat(s)} p_j \leq P_{tot}^{(s)} \right\}} \left[ \sum_{j \in Sat(s)} \mu_j(p_j, c_j) \right]$$

Decoupled Policy:

Power Alloc.: Allocate to maximize  $\sum_{j \in Sat(s)} \mu_j(p_j, c_j(t))$

Routing: Route every packet arriving in a timeslot to the queue  $i$  with the least unfinished work.

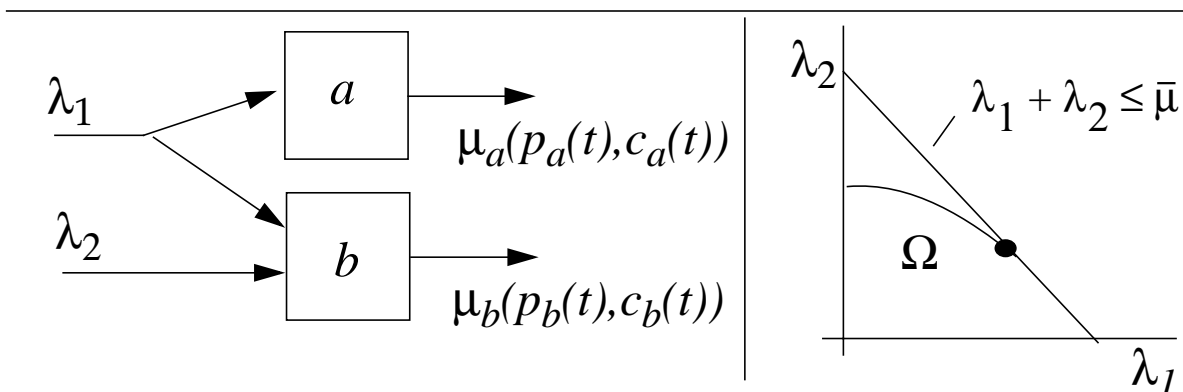


Variation of the problem: Many input types

Routing: Route packets from stream \$A\_i\$ to the shortest queue in its class \$Q\_i\$.

Power Allocation: Each Satellite Carries out the Dynamic Power Allocation Policy: Maximize:

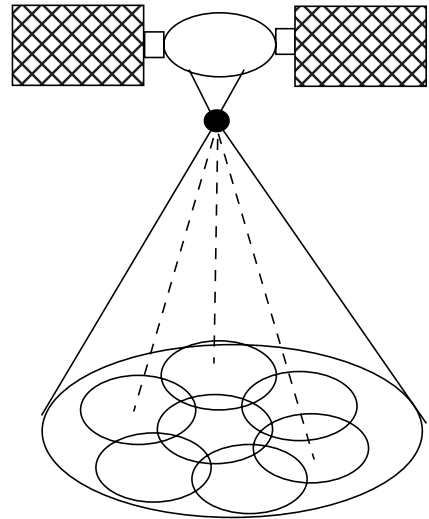
$$\sum_{j \in Sat(s)} U_i(t) \mu(p_j, c_j(t))$$



## Connectivity Constraints:

To limit channel interference, define *connectivity sets* to ensure power does not affect other channels:

$$\text{Power } \vec{P}(t) \in \{\mathcal{P}_1, \mathcal{P}_2, \dots, \mathcal{P}_R\}$$



Example -- Limited to 3 Servers:

$$\mathcal{P}_r = \left\{ (p_1, p_2, p_3, 0, \dots, 0) \in \mathbb{R}^N \mid p_j \geq 0, \sum_{j=1}^3 p_j \leq P_{tot} \right\}$$

Capacity Region:

$$\vec{\lambda} \in \Omega \triangleq \sum_{\vec{C}} \pi_{\vec{C}} \text{Convex Hull} \left( \{ \vec{\mu}(\vec{P}, \vec{C}) \mid (\vec{P} \in \mathcal{P}_r) \} \Big|_{r=1}^R \right)$$

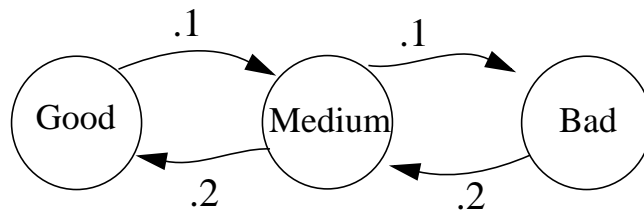
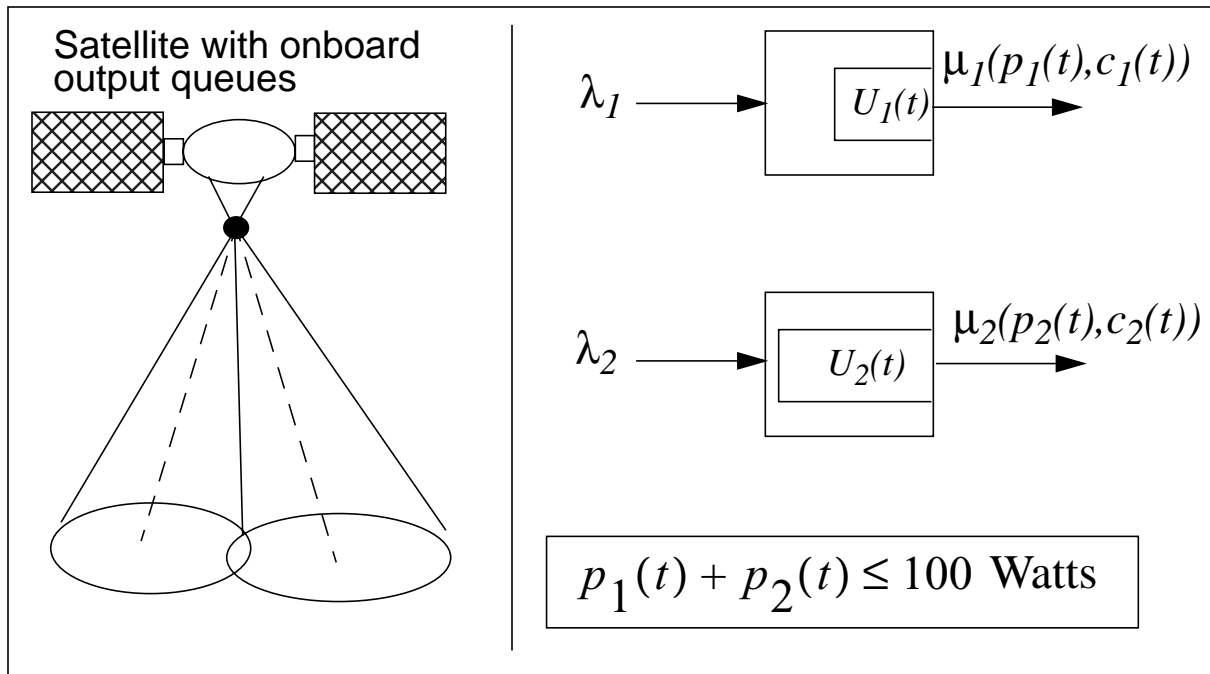
Power Allocation Policy:

$$\text{Maximize: } \sum_{j=1}^N U_j(t) \mu_j(p_j, c_j(t))$$

$$\text{Subject to: } \vec{P} \in \mathcal{P} = \{\mathcal{P}_1, \dots, \mathcal{P}_R\}$$

## Numerical and Simulation Results:

### Markovian Channel Dynamics / Poisson inputs



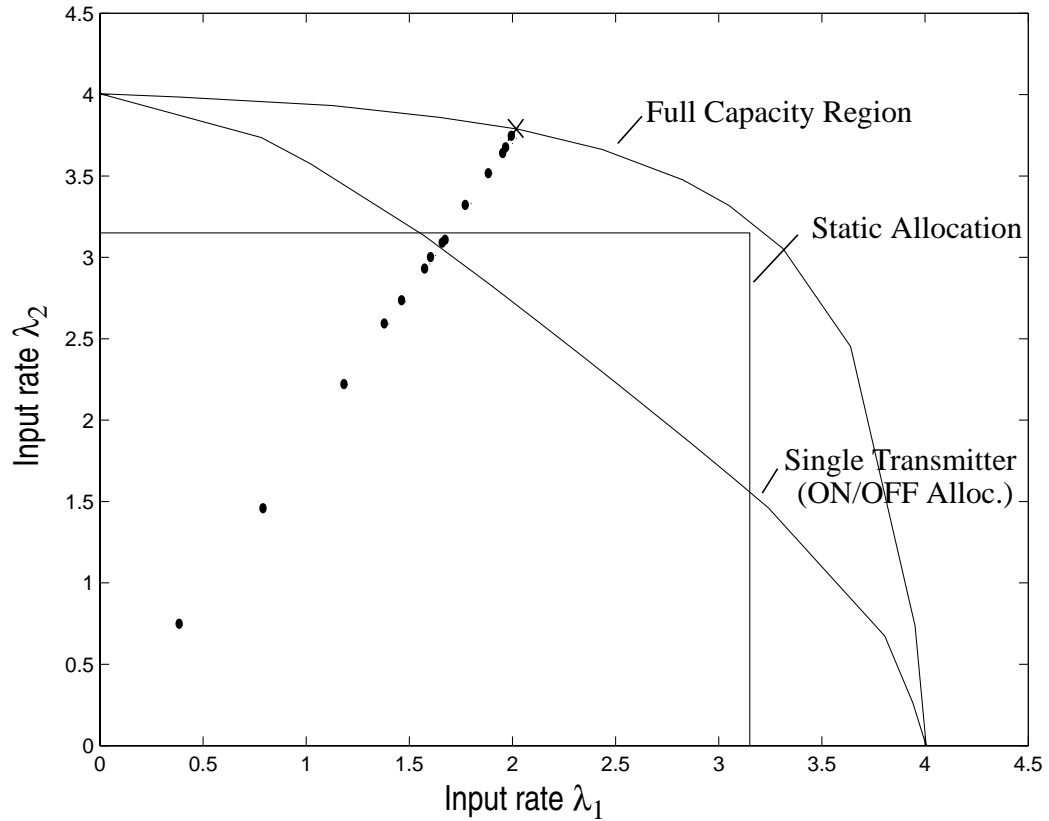
Log-normal distribution of  $\alpha_i$  for each of the three channel conditions:

	mean	variance
Good	15 db	.264 db-squared
Medium	10 db	.868 db-squared
Bad	0 db	.145 db-squared

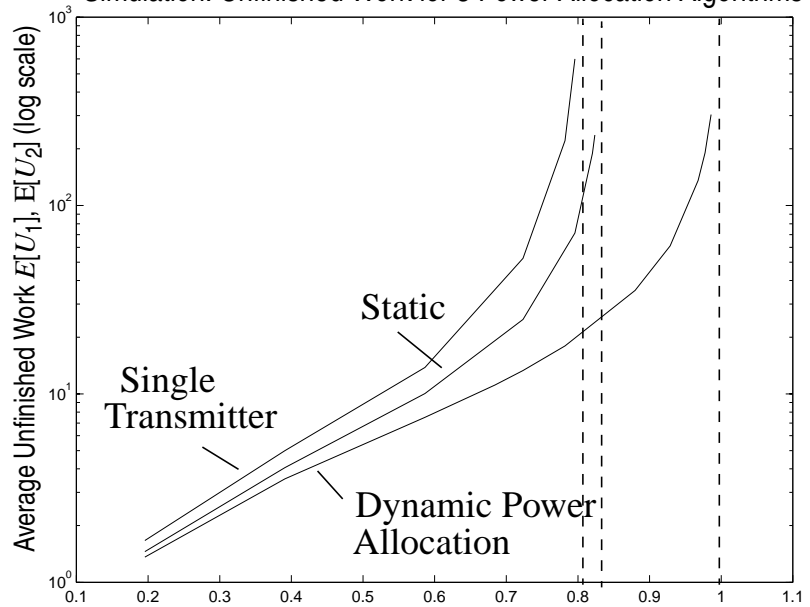
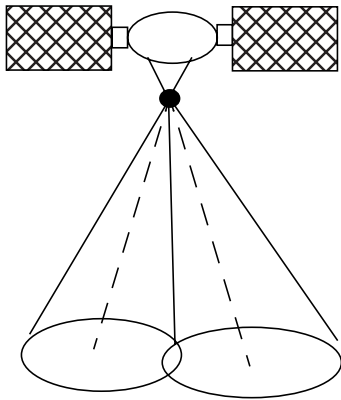


# Concluding Slide:

## Stability Regions for Three Power Allocation Algorithms



## Simulation: Unfinished Work for 3 Power Allocation Algorithms



Fraction "f" from the capacity region boundary:  $(\lambda_1, \lambda_2) = (2.05, 3.79)f$