

Lemma 1: If \underline{r}^* is the optimal solution to (P) under (3), then for any feasible rate vector \underline{r} ,

$$\sum_{\sigma \in S} \frac{r_{\sigma} - r_{\sigma}^*}{r_{\sigma}^{*\alpha}} \leq 0$$

(\underline{r}^* is called α -proportionally fair)

Proof:

$$\sum_{\ell} P_{\ell} \sum_{\sigma \in S_{\ell}} r_{\sigma} \leq \sum_{\ell} P_{\ell} C_{\ell} \quad \text{fr (1)}$$

$$= \sum_{\ell} P_{\ell} \sum_{\sigma \in S_{\ell}} r_{\sigma}^* \quad \text{fr (5)}$$

$$\Rightarrow 0 \geq \sum_{\ell} P_{\ell} \sum_{\sigma \in S_{\ell}} (r_{\sigma} - r_{\sigma}^*)$$

$$= \sum_{\sigma} \sum_{\ell \in L_{\sigma}} P_{\ell} (r_{\sigma} - r_{\sigma}^*)$$

$$= \sum_{\sigma} (r_{\sigma} - r_{\sigma}^*) \left. \frac{\partial U}{\partial r_{\sigma}} \right|_{\underline{r}^*} \quad \text{fr (4)}$$

$$= \sum_{\sigma} (r_{\sigma} - r_{\sigma}^*) (r_{\sigma}^*)^{-\alpha} \quad \text{fr (3)} \quad \square$$

to show $\alpha \rightarrow \infty$ gives the MMF allocⁿ, let $\underline{r}^*(\alpha)$ be the optimal solution to (P), & let \underline{r} be a feasible rate vector st. $\underline{r} \neq \underline{r}^*$ & for some session j , $r_j > r_j^*(\alpha)$ for $\alpha \rightarrow \infty$

• from Lemma 1,

$$\sum_{\sigma \in S} \frac{r_{\sigma} - r_{\sigma}^*(\alpha)}{(r_{\sigma}^*(\alpha))^{\alpha}} \leq 0$$

$$\Rightarrow \frac{r_j - r_j^*(\alpha)}{(r_j^*(\alpha))^{\alpha}} \leq - \sum_{\sigma \neq j} \frac{r_{\sigma} - r_{\sigma}^*(\alpha)}{(r_{\sigma}^*(\alpha))^{\alpha}}$$

$$\Rightarrow 1 \leq - \sum_{\sigma \neq j} \frac{r_{\sigma} - r_{\sigma}^*(\alpha)}{(r_{\sigma}^*(\alpha))^{\alpha}} \frac{(r_j^*(\alpha))^{\alpha}}{r_j - r_j^*(\alpha)}$$

since $r_j - r_j^*(\alpha) > 0$

$$\leq \sum_{\sigma \neq j} - \frac{r_{\sigma} - r_{\sigma}^*(\alpha)}{r_j - r_j^*(\alpha)} \left(\frac{r_j^*(\alpha)}{r_{\sigma}^*(\alpha)} \right)^{\alpha}$$

positive terms only

$$\Rightarrow r_j^*(\alpha) \geq r_k^*(\alpha) \quad \text{for } \alpha \rightarrow \infty$$

for some $k \neq j$, otherwise all RHS terms would go to 0 as $\alpha \rightarrow \infty$.

Also, since we consider only strictly positive terms on the RHS, & $r_j > r_j^*(\alpha)$ for $\alpha \rightarrow \infty$,

$$r_k < r_k^*(\alpha) \leq r_j^*(\alpha)$$

$\Rightarrow \underline{r}^*$ is MMF

Congestion control as a decentralized solution to resource allocation problem

- different types of congestion control mechanisms correspond to different utility functions & different distributed optimization algorithms
- TCP can be viewed as approximating a primal penalty function algorithm with utility function $\approx -\frac{1}{r_\sigma d_\sigma^2}$ for small RTT d_σ & small loss probability.

Primal penalty function algorithm

- Problem (P'):

$$\max_{r \geq 0} V(r) = \sum_{\sigma \in S} U_\sigma(r_\sigma) - \sum_{l \in L} \int_0^{f_l(r)} h_l(y, \gamma_l) dy$$

- $h_l(\cdot)$ is the penalty (or price) function for link l

- assumption: for each $l \in L$, $h_l(y, \gamma_l)$ is a continuous nondecreasing function of y & a decreasing function of γ_l , st.

$$\int_0^f h_l(y, \gamma_l) dy \rightarrow \infty \text{ as } f \rightarrow \infty$$

→ since we have assumed

$U_\sigma(\cdot)$ is strictly concave,

$V(\cdot)$ is strictly concave in r

- parameter γ_l is used to adapt the penalty according to congestion

- assumption: $\forall \sigma \in S, U_\sigma(r_\sigma) \rightarrow -\infty$ as $r_\sigma \rightarrow 0$

→ optimal solution satisfies $r_\sigma > 0 \quad \forall \sigma \in S$

- optimal r satisfies

$$\frac{\partial V}{\partial r_\sigma} = 0 \quad \forall \sigma \in S$$

$$\Rightarrow \frac{\partial U}{\partial r_\sigma} - \sum_{l \in L_\sigma} h_l(f_l(r), \gamma_l) = 0 \quad \forall \sigma \in S$$

p_l : link (marginal) price

q_σ : session σ route price

- decentralized iterative algorithm:

• let $p_l(t) = h_l(f_l(r(t)), \gamma_l(t))$

$$q_\sigma(t) = \sum_{l \in L_\sigma} p_l(t)$$

be the link & route prices at time t

• for each $\sigma \in S$,

$$\frac{dr_\sigma}{dt} = k_\sigma(r_\sigma) \left(\frac{\partial U}{\partial r_\sigma} - q_\sigma(t) \right)_r^+ \quad (*)$$

where $k_\sigma(r_\sigma)$ is any nondecreasing continuous function st. $k_\sigma(r_\sigma) > 0 \quad \forall r_\sigma > 0$

• for each $l \in L$,

$$\frac{d\gamma_l}{dt} = \alpha_l (C_l - f_l(r(t)))_{\gamma_l}^+$$

• where $\alpha_l > 0$.

$$*(g(r))_r^+ = \begin{cases} g(r) & \text{if } r > 0, \\ & \text{or } r = 0 \text{ \& } g(r) \geq 0 \\ 0 & \text{if } r = 0 \text{ \& } g(r) < 0 \end{cases}$$

(if $C_l < f_l(r)$, $\gamma_l \downarrow \rightarrow$ link price \uparrow)