Experiences with multi-path algorithms for joint routing and congestion control

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Image: A matrix and a matrix



Talk arises from on-going collaboration with

- Frank Kelly
- Peter Key
- Tom Voice



Image: A matrix and a matrix

Network model Algorithms

Optimization problem

$$\max_{x_r, y_s \ge 0} \sum_{s \in S} U_s(y_s)$$

subject to

$$y^q_s = \sum_{r \in s} x^q_r \qquad \forall s \in S$$

and

$$\sum_{r:j\in r} x_r \leq C_j \qquad \forall j \in J$$

where q = p/(p + 1) for some p > 0.

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Network model Algorithms

Effect of q: valuing path diversity

The quantity

$$m_q(x) = (\sum_i x_i^q)^{1/q}$$

is superadditive, that is,

$$m_q(x) + m_q(y) \leq m_q(x+y)$$

for all $x, y \ge 0$ and $0 < q \le 1$. The graph shows $m_q((\delta, 1 - \delta))$ for $0 < \delta < 1$ and varying values of q.



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Network model Algorithms

Choice of utility function

Assume

$$U_{s}(y) = w_{s} \frac{y^{1-\alpha}}{1-\alpha} \qquad \alpha \neq 1$$

and

$$U_{s}(y) = w_{s}\log(y)$$
 $\alpha = 1$

with $w_s > 0$ and $\alpha > 0$.



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Network model Algorithms

Reduced splitting dual algorithm

For each route r,

$$\mathbf{x}_r(t) = \lambda_r(t)^{-(p+1)} \mathbf{w}_{\mathbf{s}(r)}^{p+1} \mathbf{y}_{\mathbf{s}(r)}(t)^{1-\alpha(p+1)}$$

where $\lambda_r(t) = \sum_{j \in r} \mu_j(t - T_{jr})$. For each link *j*,

$$\dot{\mu}_j(t) = \kappa_j \mu_j(t) (\mathbf{z}_j(t) - \mathbf{C}_j)^+_{\mu_j(t)}$$

where $z_j(t) = \sum_{r:j \in r} x_r(t - T_{rj})$. For each source *s*,

$$\dot{y}_{s}(t) = \kappa_{s} y_{s}(t)^{1/p} \left(\sum_{r \in s} x_{r}(t - T_{r})^{q} - y_{s}(t)^{q} \right)^{+}_{y_{s}(t)}$$

(cf. T. Voice (2005) Stability of multi-path dual algorithms for joint routing and flow control.)



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Example 1 Example 2 Example 3

Network and user parameters

- sources, $s \in S$ comprising one or more routes $r \in R$
- routes, *r* comprising one or more resources, $j \in r$
- resource capacity C_j , $j \in J$
- delays $T_r = T_{rj} + T_{jr}$
- q = p/(p+1)
- $w_s > 0$
- α_s = 1

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Example 1 Example 2 Example 3

Algorithm parameters

- κ_j for each resource $j \in J$
- κ_s for each source $s \in S$
- initial prices, $\mu_j(0)$, for $j \in J$
- initial flows, $y_s(0)$, for $s \in S$

Image: A matrix and a matrix

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Example 1 Example 2 Example 3

Example 4-node network — case 1



Gibbens Multi-path algorithms

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Example 1 Example 2 Example 3

Results 1



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Example 1 Example 2 Example 3

Example 4-node network — case 2



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Example 1 Example 2 Example 3

Results 2(a)



Example 1 Example 2 Example 3

Results 2(b)



Numerical results

Example 2

Results 2(c)



Example 1 Example 2 Example 3

Example 4-node network — case 3



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Example 1 Example 2 Example 3

Results 3



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Incentives Wireless Route selection

Extensions

Potential extensions to and connections with the following

- Incentives for resource sharing in ad-hoc networks
- Integration with wireless, not wired, networks
- Adaptive route selection algorithms

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Incentives Wireless Route selection

Incentives for resource sharing

Build on approach taken in:

 Jon Crowcroft, Richard Gibbens, Frank Kelly and Sven Östring Modelling incentives for collaboration in mobile ad hoc networks
Performance Evaluation (2004) 57, 427–439



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Incentives Wireless Route selection

Credit transfers between nodes

- Arriving users join with unit credit
- Credit transfers determined by the congestion prices μ_j are exchanged between sources and the nodes used along routes
- Departing user takes their credit with them
- Concurrently, adjust each user's credit balance over time towards unity.

Cooperation provided by means of congestion fees paid by users from credit received from other users for forwarding via their own congested resources.



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Incentives Wireless Route selection

The form of capacity constraints in wireless networks

Mung Chiang

Balancing transport and physical layers in wireless multihop networks: jointly optimal congestion control and power control

IEEE JSAC, Vol 23, No 1, January 2005

$$\max_{x,P\geq 0}\sum_{s}U_{s}(x_{s})$$

subject to

$$\sum x_{s} \leq c_{\ell}(P) \qquad orall \ell$$

where

$$c_{\ell}(P) = \frac{1}{T} \log(1 + KSIR_{\ell}(P)) \quad \text{and} \quad SIR_{\ell}(P) = \frac{P_{\ell}G_{\ell\ell}}{\sum_{k \neq \ell} P_k G_{\ell k} + \eta_{\ell}}$$

Incentives Wireless Route selection

Adaptive route selection algorithms

Basic question

How do we dynamically adjust the active set of routes in the face of varying congestion and mobility?

Use the behaviour of the multi-path congestion/flow control algorithm as the feedback signal within a sticky random algorithm, much as for circuit-switched networks

R.J. Gibbens, F.P. Kelly and P.B. Key Dynamic alternative routing — modelling and behaviour In *Twelfth International Teletraffic Congress*. North-Holland (1988), Turin.



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Summary

- Algorithms have been implemented and behaviour explored in example networks
- Helps build an understanding of the interactions between the many issues and parameters
- Potential extensions to:
 - incentives for collaboration in mobile ad hoc networks
 - alternative capacity constraints such as those for wireless
 - algorithms for dynamic route set selection