

Introduction to Routing and Congestion Control

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Outline

1 Introduction

2 Convex Functions and Optimization

3 Routing Problems

Overview

- The role of routing: decide which routes through the network should be used to get the data from sources to destinations.
- The role of congestion control: determine the data rates for the users so that fairness among users is achieved, and a reasonable operating point is found on the tradeoffs between throughput, delay or loss.

Introduction

- Suppose that Ω is a subset of \mathbb{R}^n for some $n \geq 1$ and Ω is a **convex set**.
- A **convex set**: any pair of points in Ω , the line segment connecting them is also in Ω , or whenever $x, x' \in \Omega$, then $ax + (1-a)x' \in \Omega$ for $0 \leq a \leq 1$.
- A function f on Ω is a *convex function* if along any line segment in Ω , f is less than or equal to the value of the linear function agreeing with f at the endpoints, or

$$f(ax + (1-a)x') \leq af(x) + (1-a)f(x')$$

for $x, x' \in \Omega$ and $0 \leq a \leq 1$.

- A **concave function** f is a function f such that $-f$ is a convex function.
- Minimizing a convex function can be translate to maximizing a concave function.

Definitions

- The gradient, ∇f , of a function f , is a vector valued function as:

$$\nabla f = \begin{pmatrix} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \\ \vdots \\ \frac{\partial f}{\partial x_n} \end{pmatrix}$$

We assume ∇f exists and continuous.

- If $x \in \Omega$ and if v is a vector such that $x + \epsilon v \in \Omega$ for small enough $\epsilon > 0$, then the **directional derivative** of f at x in the direction of v :

$$\lim_{\epsilon \rightarrow 0} \frac{f(x + \epsilon v) - f(x)}{\epsilon} = (\nabla f(x)) \cdot v = \sum_i \frac{\partial f(x)}{\partial x_i} v_i$$

- We have $x^* \in \operatorname{argmin}_{x \in \Omega} f(x)$ if x^* minimizes f over Ω .

Proposition

Suppose f is a convex function on a convex set Ω with a continuous gradient function ∇f . Then $x^* \in \operatorname{argmin}_{x \in \Omega} f(x)$ if and only if $(\nabla f(x^*)) \cdot (x - x^*) \geq 0$ for all $x \in \Omega$.

Model

The flow rates of users are taken as given, the problem is to determine which routes should be taken. We define

- J is the set of links.
- R is the set of routes. Each route r is associated with a subset of links. We write $j \in r$ to denote link j is in route r . In general, we allow two different routes to use the same set of links.
- A is the link-route incidence matrix with $A_{j,r} = 1$ if $j \in r$ and $A_{j,r} = 0$ otherwise.
- S is the set of users. Each user s is associated with a subset of the set of routes. We write $r \in s$ to denote that route r serves user s .
- H is the user-route incidence matrix, $H_{s,r} = 1$ if $r \in s$ and $H_{s,r} = 0$ otherwise.

Model

- y_r is the flow on route r .
- f_j is the flow on link j : $f_j = \sum_{r:j \in r} y_r$, or in vector form, $f = Ay$.
- x_s is the total flow available to user s : $x_s = \sum_{r \in S} y_r$, or in vector form, $x = Hy$. The vector $x = (x_s : s \in S)$ is fixed for the routing problem, and variable in the joint congestion control and routing problem.
- $D_j(f_j)$: is the cost of carrying flow f_j on link j . The function D_j is assumed to be a convex function, continuous differentiable, and increase function \mathbf{R}_+ . The cost associated with flow rate f_j on link j is $D_j(f_j)$.

The Routing Problem

$ROUTE(x, H, A, D)$:

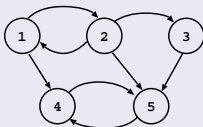
$$\min F(y) = \sum_{j \in J} D_j \left(\sum_{r: j \in r} y_r \right)$$

subject to

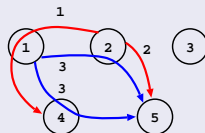
$$x = Hy, \quad \text{over } y \geq 0.$$

- We say that vector y is a **feasible flow** meeting demand x if $y \geq 0$ and $x = Hy$.
- A possible choice of D_j is $D_j(f_j) = \frac{f_j}{C_j - f_j} + f_j d_j$ where C_j is the capacity of link j and d_j is the propagation delay of link j .
- If we use an $M/M/1$ model, $D_j(f_j)$ represents the mean number of packets in transit of link j . So the total cost $F(y)$ represents the mean number of packets in transits in the network. Then $F(y) / \sum_s x_s$ is the average delay of packets entering the network.

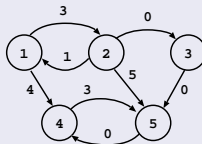
Example



(a): network



(b): route flows



(c): link flows

- Set of links $J = \{(1, 2), (1, 4), (2, 1), (2, 3), (2, 5), (3, 5), (4, 5), (5, 4)\}$.
- Set of users $S = \{(1, 5), (2, 4), (2, 5)\}$
- We assume user $(1, 5)$ is served by routes $(1, 2, 5), (1, 4, 5), (1, 2, 3, 5)$, user $(2, 4)$ is served by routes $(2, 1, 4), (2, 5, 4)$, user $(2, 5)$ is served by routes $(2, 5), (2, 3, 5)$.

Example: continue

- Flow rates on each route: $y_{(1,2,5)} = 3.0$, $y_{(1,4,5)} = 3.0$,
 $y_{(1,2,3,5)} = 0$, $y_{(2,1,4)} = 1.0$, $y_{(2,5,4)} = 0$, $y_{(2,5)} = 2.0$, $y_{(2,3,5)} = 0$.
- This implies that flow rates for each source: $x_{(1,5)} = 6.0$,
 $x_{(2,4)} = 1.0$, $x_{(2,5)} = 2.0$.
- The link rates: $f_{(1,2)} = 3$, $f_{(1,4)} = 4$, $f_{(2,1)} = 1$, ...etc, as seen in the figure.
- The result cost is:

$$F(y) = D_{1,2}(3) + D_{1,4}(4) + D_{2,1}(1) + D_{2,3}(0) + D_{2,5}(5) + D_{3,5}(0) + D_{4,5}(3) + D_{5,4}(0).$$

- Considering changing the flow rates on a route. If $y_{1,2,5}$ were increase to $y_{1,2,5} + \epsilon$ for some $\epsilon > 0$, then the cost $F(y)$ would increase by:

$$\begin{aligned} \Delta F &= D_{1,2}(3 + \epsilon) - D_{1,2}(3) + D_{2,5}(5 + \epsilon) - D_{2,5}(5) \\ &= \epsilon(D'_{1,2}(3) - D'_{2,5}(5)) + O(\epsilon). \end{aligned}$$

Example: continue

- Similarly, if we decrease $y_{1,4,5}$ by ϵ , the cost would decrease by

$$\epsilon(D'_{1,4}(4) - D'_{4,5}(3)) + O(\epsilon).$$

- If $y_{1,2,5}$ were increased by ϵ and if $y_{1,4,5}$ were simultaneously decreased by ϵ (of ϵ of the flow were *deviated* to route (1, 2, 5) from route (1, 4, 5)), the *flow deviation cost* would be:

$$\epsilon(D'_{1,2}(3) + D'_{2,5}(5) - D'_{1,4}(4) - D'_{4,5}(3)) + O(\epsilon)$$

- first order approximation, the change in cost due to the flow deviation is ϵ times the first derivative length $D'_{1,2}(3) + D'_{2,5}(5)$ of route (1, 2, 5) minus the first derivative length $D'_{1,4}(4) + D'_{4,5}(3)$ of route (1, 4, 5).
- The flow deviation will decrease the cost if flow is deviated from a *longer* route to a *shorter* route, where length of a route is the sum of the first derivative link lengths, $D_j(f_j)$.

Important Observation

- The above example implies that a *necessary condition* for y to be optimal is that for any s and any $r \in s$, the rate y_r is strictly positive only if r has the **minimum first derivative length** among the routes serving s .
- Since
 - the set of feasible rate vectors j meeting the demand x is a convex set,
 - the cost function F is convex,
 - It follows from convex optimization that the first order necessary condition for optimality is also sufficient for optimality.

Proposition

A feasible flow $\mathbf{y} = (y_r : r \in R)$ meeting demand \mathbf{x} minimizes the cost over all such flows, if and only if there exists a vector $(\lambda_s : s \in S)$ so that $\sum_{j \in r} D'_j(f_j) \geq \lambda_s$, with *equality* if $y_r > 0$, whenever $r \in s \in S$.

Comment

Given a flow vector \mathbf{y} , let δ_r denote the first derivative length of route r , or $\delta_r = \sum_{j \in r} D'_j(\sum_{r': j \in r'} y_{r'})$. If the rate requirement of a user s is $x_s > 0$, then the parameter λ_s must satisfy $\lambda_s = \min_{r \in S} \delta_r$.

The Flow Deviation Algorithm

One iteration is described. At the beginning of an iteration, a feasible flow vector \mathbf{y} is given. We find the next such vector $\bar{\mathbf{y}}$.

- Compute the link flow f_j and first derivative lengths $D_j(f_j)$ for all links $j \in J$.
- For each user s , find a route r_s^* serving s with the minimum first derivative link length.
- Let $\alpha \in [0, 1]$. For each user s , deviate a fraction α of the flow from each of the other routes serving s to r_s^* , that is

$$\bar{y}_r = \begin{cases} (1 - \alpha)y_r & \text{if } r \in s \text{ and } r \neq r_s^* \\ y_r + \alpha \left(\sum_{r': r' \neq r' \in s} y_{r'} \right) & \text{if } r = r_s^* \end{cases}$$

Adjust α to minimize $F(\bar{\mathbf{y}})$. The resulting $\bar{\mathbf{y}}$ is the solution of one iteration.