## **Network Optimization**

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4 Abstract

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## 6 1 Network Traffic Problem

In this section, we consider the problem of optimizing network traffic. A network is a graph G=(V,E) with vertices  $v_i\in V$  and edges  $e_{ij}\in E=V\times V$ . We would like to maximize the total traffic through a series of K paths across the network. A path is a sequence of edges  $p_{st}^k=(e_{sv_1},e_{v_1v_2},\ldots,e_{v_mt})$  from source vertex s to target vertex t. A path contributes a constant amount of traffic  $x_p$  to each included edge  $e\in p$ . We denote the set of paths that pass through a particular edge by  $\pi(e)=\{p\mid e\in p\}$ . We additionally have the following constraints: traffic must be nonnegative  $x_p\geq 0$ , and each edge has a capacity constraint so that the total traffic on that edge cannot exceed  $c_e\geq \sum_{p\in\pi(e)} x_p$ .

This yields the following optimization problem:

$$\begin{array}{ll} \text{maximize} & \sum_{p} x_{p} \text{ or } \sum_{e} \sum_{p \in \pi(e)} x_{p} \\ \\ \text{subject to} & \sum_{p \in \pi(e)} x_{p} \leq c_{e}, \forall e \\ \\ & x_{p} \geq 0, \forall p. \end{array} \tag{1}$$

The first objective assigns equal weight to each path, while the second objective weights paths based on length. We focus on the first objective, since the goal is to maximize the amount of traffic each customer receives (represented as a path) rather than the congestion of the network.

While this form is compact, we would like to apply the ADMM algorithm, which applies to problems of the form

minimize 
$$f(x) + g(z)$$
  
subject to  $Ax + Bz = c$ . (2)

To massage the problem in Eqn. 1 into the correct form, we will introduce two sets of new variables: slack variables  $s_e$  for each edge<sup>1</sup> and decoupled edge weights  $z_{pe}$  for each edge and path. The new problem is given by

minimize 
$$f(x) + g(z) + h(s)$$
 subject to 
$$c_e - \sum_{p \in \pi(e)} z_{pe} - s_e = 0, \forall e$$
 
$$x_p - z_{pe} = 0, \forall e, \forall p \in \pi(e)$$
 
$$x, s, z \succ 0,$$
 (3)

where  $f(x) = -\sum_p x_p$ ,  $g(z) = \sum_{pe} \bar{\delta}(z_{pe} \ge 0)$ ,  $h(s) = \sum_e \bar{\delta}(s_e \ge 0)$ . We use the delta notation to indicate a function

 $\bar{\delta}(b) = \begin{cases} \infty & \text{if condition } b \text{ does not hold,} \\ 0 & \text{o.w.} \end{cases}$ 

This problem is decomposable along each path (for x) or edge (for z and s) as follows:

minimize 
$$-\sum_{p} x_{p} + \sum_{e} \sum_{p \in \pi(e)} \bar{\delta}(z_{pe}) + \sum_{e} \bar{\delta}(s_{e})$$
subject to 
$$c_{e} - \sum_{p \in \pi(e)} z_{pe} - s_{e} = 0, \forall e$$

$$x_{p} - z_{pe} = 0, \forall e, \forall p \in \pi(e)$$

$$x, s, z \succeq 0,$$

$$(4)$$

where the summation in the objective is nested.

The augmented Lagrangian for this problem is

$$\mathcal{L}_{\phi}(x, z, s, \lambda) = -\sum_{p} x_{p} + \sum_{p} \sum_{e \in p} \bar{\delta}(z_{pe}) + \sum_{e} \bar{\delta}(s_{e}) + \lambda^{\top} F(x, z, s) + (\rho/2) \|F(x, z, s)\|_{2}^{2},$$
 (5)

where

$$F(x, z, s) = \begin{bmatrix} F_1 \\ F_2 \\ F_3 \\ F_4 \\ F_5 \end{bmatrix}$$

and

$$[F_1]_e = c_e - \sum_{p \in \pi(e)} z_{pe} - s_e, [F_2]_{pe} = x_p - z_{pe}, [F_3]_p = x_p, [F_4]_e = s_e, [F_5]_{pe} = z_{pe}.$$

The elements of  $\lambda$  corresponding to  $F_1$  are  $\lambda_{1,e}$ ,  $F_2$  are  $\lambda_{2,pe}$ , and so forth.

<sup>&</sup>lt;sup>1</sup> The slack variable replace the inequality constraints with equality constraints.

The ADMM updates are as follows:

$$x^{k+1} := \underset{x}{\operatorname{argmin}} \mathcal{L}_{\rho}(x, z^{k}, s^{k}, \lambda^{k})$$

$$z^{k+1} := \underset{z}{\operatorname{argmin}} \mathcal{L}_{\rho}(x^{k+1}, z, s^{k}, \lambda^{k})$$

$$s^{k+1} := \underset{s}{\operatorname{argmin}} \mathcal{L}_{\rho}(x^{k+1}, z^{k+1}, s, \lambda^{k})$$

$$\lambda^{k+1} := \lambda^{k} + \rho(F(x^{k+1}, z^{k+1}, s^{k+1})).$$
(6)

We can compute  $\operatorname{argmin}_x \mathcal{L}_{\rho}(x, z^k, s^k, \lambda^k)$  by restricting our attention to terms of  $\mathcal{L}_{\rho}$  involving x and setting the derivative equal to 0:

$$0 = \nabla_{x_p} (-x_p + \sum_{e \in p} \lambda_{2,pe} (x_p - z_{pe}) + \lambda_{3,p} x_p + (\rho/2) (x_p^2 + \sum_{e \in p} (x_p - z_{pe})^2))$$

$$= -1 + \sum_{e \in p} \lambda_{2,pe} + \lambda_{3,p} + \rho x_p + \rho \sum_{e \in p} (x_p - z_{pe})$$

$$x_p = \max(0, \frac{1 - \sum_{e \in p} \lambda_{2,pe} - \lambda_{3,p} + \rho \sum_{e \in p} z_{pe}}{(1 + |p|)\rho}).$$

We perform a similar computation for  $\operatorname{argmin}_z \mathcal{L}_{\rho}(x^{k+1},z,s^k,\lambda^k)$ :

$$0 = \nabla_{z_{pe}} \bar{\delta}(z_{pe}) + \lambda_{1,e} (c_e - \sum_{p \in \pi(e)} z_{pe} - s_e) + \lambda_{2,pe} (x_p - z_{pe}) + \lambda_{5,pe} z_{pe}$$

$$+ (\rho/2) ((c_e - \sum_{p \in \pi(e)} z_{pe} - s_e)^2 + (x_p - z_{pe})^2 + z_{pe}^2)$$

$$= -\lambda_{1,e} - \lambda_{2,pe} + \lambda_{5,pe} + \rho (-c_e + \sum_{p \in \pi(e)} z_{pe} + s_e + 2z_{pe} - x_p).$$

This gives us a system of equations for each edge  $0 = A_e z_e + b_e$  allowing us to solve for  $z_e = -A_e^{-1}b_e$ . Let  $P_e = |\pi(e)|$ . We then have

$$A_e = \mathbf{1}_{P_e \times P_e} + 2I_{P_e \times P_e}$$
$$[b_e]_p = (-\lambda_{1,e} - \lambda_{2,pe} + \lambda_{5,pe} + \rho(-c_e + s_e - x_p).$$

Then, for  $\operatorname{argmin}_s \mathcal{L}_{\rho}(x^{k+1}, z^{k+1}, s, y^k)$ :

$$0 = \nabla_{s_e} \bar{\delta}(s_e) + \lambda_{1,e} (c_e - \sum_{p \in \pi(e)} z_{pe} - s_e) + \lambda_{4,e} s_e + (\rho/2) ((c_e - \sum_{p \in \pi(e)} z_{pe} - s_e)^2 + s_e^2)$$

$$= -\lambda_{1,e} + \lambda_{4,e} + \rho(-c_e + \sum_{p \in \pi(e)} z_{pe} + 2s_e)$$

$$s_e = \max(0, \frac{\lambda_{1,e} - \lambda_{4,e} + \rho(c_e - \sum_{p \in \pi(e)} z_{pe})}{2\rho}).$$