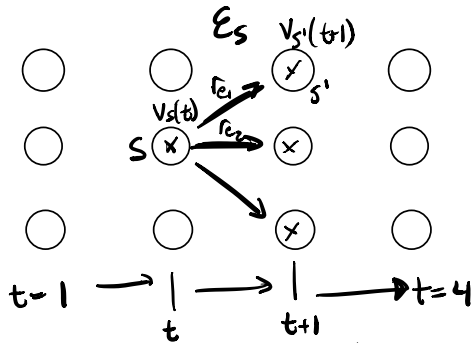


NETWORK FLOW PROBLEMS

- SHORTEST PATH LP (QP, CONVEX)
- MARKOV DECISION PROCESSES LP ROUTING GAMES

Dynamic Programming • "Cost-to-go"
(Reward-to-go)
Value function

Deterministic



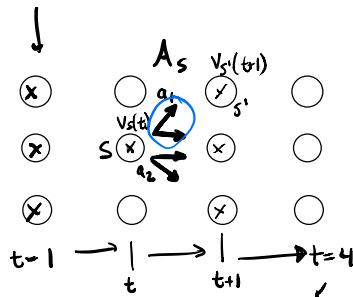
Bellman Eqn:

$$V_s(t) = \max_{e \in E_s} \{ r_e + V_{s'}(t+1) \}$$

$e: s \rightarrow s'$

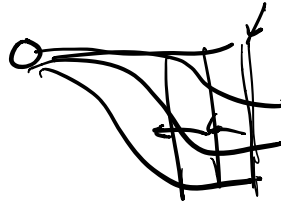
$$e_{opt}(t) \in E_s = \arg \max$$

Stochastic

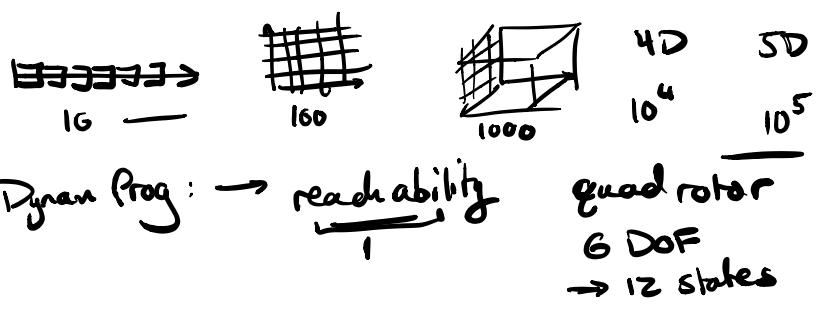


$$V_s(t) = \max_{a \in A_s} \left\{ r_a^t + \sum_{s'} P(s'|s,a) V_{s'}(t+1) \right\}$$

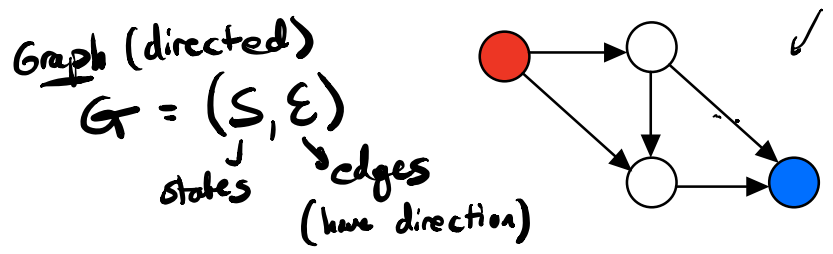
$P(s'|s,a)$ prob of transitioning to s' if you take action a in state s



Model our position as a prob. distribution over the states



SHORTEST PATH PROBLEMS:



$|S|$: # of states $|E|$: # of edges

- : $s_0 \in S$ origin node
- : $s_d \in S$ destination node

Ea. edge has a travel cost: $C_e \in \mathbb{R}_+$

GRAPH MATRICES:

Incidence Matrices

$E_0 \in \mathbb{R}^{|S| \times |E|}$ $[E_0]_{se} = \begin{cases} 1; & \text{if } e \text{ originates at } s \\ 0; & \text{otherwise} \end{cases}$

$E_i \in \mathbb{R}^{|S| \times |E|}$ $[E_i]_{se} = \begin{cases} 1; & \text{if } e \text{ terminates in } s \\ 0; & \text{otherwise} \end{cases}$

$E = E_i - E_0$

\hookrightarrow "node edge incidence matrix."

SIDE NOTE: Laplacian $L = EE^T \in \mathbb{R}^{|S| \times |S|} \rightarrow$ Gramian of E

eigenvalues of L , consensus dynamics
 "how does information spread through graph"

$\dot{x} = -Lx$
 Laplacian: "shape of a graph"
 $(EE^T)^{1/2}$: shape of a graph

$E = \underbrace{(EE^T)^{1/2}}_{\text{shape of rows}} \underbrace{(EE^T)^{-1/2}}_{\text{rows are orthonormal}} E$

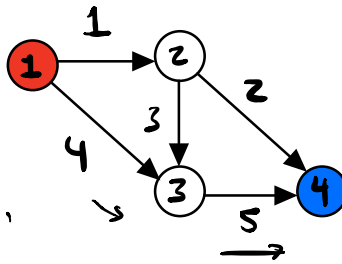
SVD of $E = U\Sigma V^T$

$(EE^T)^{1/2} = U\Sigma U^T$ (positive def.)
 $(EE^T)^{-1/2} E = UV^T$ (rotation)

$E = 4 \begin{bmatrix} -1 & 0 & 0 & -1 & 0 \\ 1 & -1 & -1 & 0 & 0 \\ 0 & 0 & 1 & 1 & -1 \\ 0 & 1 & 0 & 0 & 1 \end{bmatrix}$

always not full row rank...

$\mathbb{1}^T E = 0^T$ (vec 1's) (vec 0's)
 $\mathbb{1}^T E_0 = \mathbb{1}^T$
 $\mathbb{1}^T E_i = \mathbb{1}^T$



$x \in \mathbb{R}_+^{|E|}$ mass flow on edges

$Ex = 0$: conservation of mass at ea. state \Leftarrow

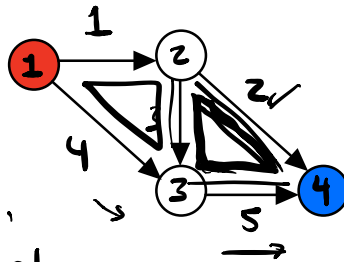
Ex. $[0 \ 0 \ 1 \ 1 \ -1]x$

$x_3 + x_4 - x_5 = 0 \Rightarrow x_5 = x_3 + x_4$

$Ex = 0 \Rightarrow$

x is a linear combination of cyclic flows

C : set of cycles of graph



cycle: directed flow around a loop of the graph

Constructed a basis for nullspace of E

$$E C = 0 \quad \text{span}(C) = \mathcal{N}(E)$$

→ the cols of C will be indicator vectors for cycles of the graph..

$$C \in \mathbb{R}^{|E| \times |C|} \quad |C|_{ec} = \begin{cases} 1 & \text{if edge } e \text{ is in cycle } c \\ & \text{and points w/ the cycle} \\ -1 & \text{if } e \text{ is in cycle } c \\ & \text{and points against the cycle} \\ 0 & \text{otherwise} \end{cases}$$

Construction of a cycle indicator:

Before: $A = [A_1 \ A_2]$
lin ind cols lin dep cols

$$A = A_1 [I \ B]$$

coeffs of A_2 wrt. A_1

$$= [A_1 \ A_1 B]$$

A_2

$$E = |S| [E_1 \ E_2]$$

lin ind.

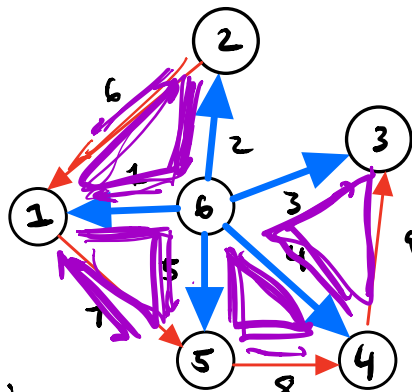
spanning tree

$$= E_1 [I \ C]$$

coeffs of the cols of E_2 w.r.t. E_1

$$E_2 = E_1 C$$

||



→ spanning: set of edges that touch ea. node
 tree: no cycles

C

$$E = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 & 1 & -1 & 0 & 0 \\ 0 & 0 & 0 & -1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & -1 & & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & -1 & 0 & 0 & 1 & -1 \\ -1 & -1 & -1 & -1 & -1 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

Every edge in E_2 corresponds to the one missing from a cycle in the graph

$$= \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & -1 & 0 & 0 \\ 0 & 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & 0 & -1 \\ -1 & -1 & -1 & -1 & -1 \end{bmatrix} \begin{bmatrix} I & C \end{bmatrix}$$

each col is identified w a cycle

$$\begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \\ 0 \\ -1 \end{bmatrix} - \begin{bmatrix} 0 \\ -1 \\ 0 \\ 0 \\ 0 \\ -1 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} \quad \begin{bmatrix} -1 \\ \vdots \\ -1 \end{bmatrix} + \begin{bmatrix} 1 \\ \vdots \\ -1 \end{bmatrix} = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$$

$$E = E_1 [I \ C]$$

basis for nullspace of $E \Rightarrow$
 ea. col of $\begin{bmatrix} C \\ -I \end{bmatrix}$ is
 an indicator
 vector for a cycle

$\begin{bmatrix} C \\ -I \end{bmatrix}$
 cols are basis
 for a nullspace

$$C = \begin{bmatrix} C \\ -I \end{bmatrix}$$

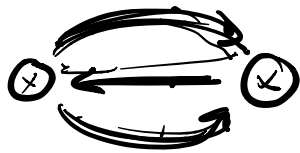
$$\{x \mid Ex = 0\} = \{x \mid x = \begin{bmatrix} C \\ -I \end{bmatrix} z\}$$

Source - Sink vector
(Origin - destination) $b \in \mathbb{R}^{|S|}$

$$b_s = \begin{cases} -1 & \text{if } s \text{ source / origin} \\ 1 & \text{if } s \text{ sink / dest.} \\ 0 & \text{otherwise} \end{cases} \leftarrow$$

\rightarrow $\left[\begin{array}{l} \underline{E} \underline{x} = \underline{b} m \\ \text{ea. row is mass conservation at a node} \end{array} \right] \leftarrow$ set of edge flows from the source to the sink (+ cyclic flows)

m : total amount of mass through network



SHORTEST PATH LP:

cost of traveling on edge $e \in E$ is $c_e > 0$

$$\left[\begin{array}{l} \min_{x \in \mathbb{R}^{|E|}} \sum_e c_e x_e = c^T x \\ \text{s.t. } \underline{E} x = \underline{b} \end{array} \right]$$

$$c \in \mathbb{R}_+^{|E|}$$

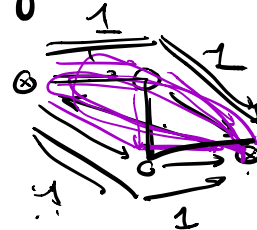
\rightarrow conservation of mass from source to sink

$$x \geq 0$$

ensures flow follows direction of edges

$$m = 1$$

$x \in \mathbb{R}^{|E|}$: indicates which edges we travel on.



$\underline{E} x = \underline{b} \leftarrow$ guarantees all mass flows from source to sink (can't just pick random edges)

$$X: \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix} \rightarrow \text{one route} \quad \begin{pmatrix} 0.5 \\ 0.5 \\ \vdots \end{pmatrix} + \begin{pmatrix} 0.5 \\ \vdots \\ 0.5 \end{pmatrix} \rightarrow 2 \text{ routes} \quad \begin{pmatrix} 0.3 \\ 1.3 \end{pmatrix} + \begin{pmatrix} 0.7 \\ 0.7 \end{pmatrix}$$

$$\min_x C^T x \quad \mathbb{1}^T x = 1$$

$$\text{s.t. } \underline{E}x = b, x \geq 0$$

$$\text{dual } v \in \mathbb{R}^{|\mathcal{E}|} \quad \mu = \mathbb{R}_+^{|\mathcal{E}|}$$

OPTIMALITY (KKT)

$$\text{Stationarity: } C^T + v^T E - \mu^T = 0 \quad \checkmark$$

$$\text{Feasibility: } Ex = b \quad \leftarrow$$

$$\text{Positivity: } x \geq 0, \mu \geq 0 \quad \leftarrow$$

$$\text{Slackness: } \underline{\mu}^T x = 0 \quad (\mu_e x_e = 0) \quad |$$

$$\frac{\partial L}{\partial x} = 0$$

$$\frac{\partial L}{\partial v} = 0 \quad \frac{\partial L}{\partial \mu} \geq 0$$

$$\text{Lagrangian: } \mathcal{L}(x, v, \mu) = C^T x + v^T (Ex - b) - \mu^T x$$

Intuition:

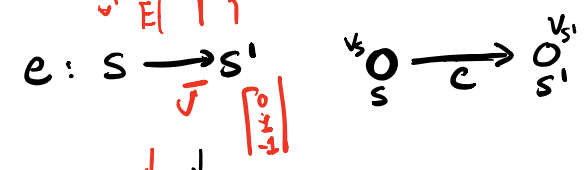
C : cost vector

x : edge flow

v : value func.

μ_e : ineff. of edge e

$$C^T + v^T E - \mu^T = 0 \quad | \text{ one eqn for ea. edge}$$



$$\rightarrow \underline{C}_e + \underline{v}_{s'} - \underline{v}_s - \underline{\mu}_e = 0 \quad \forall e \in \mathcal{E}$$

Sum this eqn over edges in a route from origin to dest.

route: $r \subset \mathcal{E}$

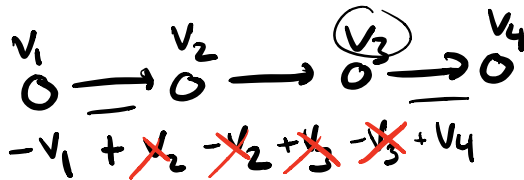
$$\Leftrightarrow v_s = C_e + v_{s'} - \mu_e \rightarrow \text{ineff.}$$

\downarrow cost on edge c \downarrow cost-to-go from s'

$$\sum_{e \in r} (C_e + v_{s'} - v_s - \mu_e) = 0$$

$$\sum_{e \in r} C_e + \sum_{e \in r} v_{s'} - v_s - \sum_{e \in r} \mu_e = 0$$

$$\left[\underbrace{\sum_{e \in r} c_e}_{\text{total travel cost}} + \underbrace{v_d - v_o}_{\text{diff in cost to go from origin to dest.}} - \underbrace{\sum_{e \in r} \mu_e}_{\text{sum of ineqs.}} = 0$$



Complementary slackness

$$\mu_e x_e = 0$$

$$x_e \geq 0$$

$$\left[\underbrace{\sum_{e \in r} c_e}_{\text{total travel cost}} = \underbrace{v_o - v_d}_{\text{cost to go from origin}} + \underbrace{\sum_{e \in r} \mu_e}_{\text{sum of ineqs.}} \right]$$

for an optimal route r^* s.t. $x_e^* > 0$ for all $e \in r^*$

$$\mu_e x_e^* = 0 \quad x_e^* > 0 \Rightarrow \underline{\mu_e = 0}$$

$$\sum_{e \in r^*} c_e = \underline{v_o - v_d} + \sum_{e \in r^*} \mu_e$$

suboptimal route r s.t. $x_e = 0$ for some $e \in r$

$$\sum_{e \in r} c_e = \underline{v_o - v_d} + \sum_{e \in r} \mu_e \quad \text{for } x_e = 0$$

μ_e can be greater than 0

$$C^T + V^T E - \mu^T = 0 \iff$$

take feasible flow x . $\Rightarrow x \geq 0$ $\boxed{Ex = b}$

$$\left[\begin{array}{l} C^T x + V^T E x - \mu^T x \\ C^T x + \underline{V^T b} - \mu^T x \end{array} \Rightarrow \underline{C^T x} = \underline{-V^T b} + \underline{\mu^T x} \right]$$

$\sum_{e \in r} c_e x_e \quad v_o - v_d \quad 0 = 0$

PICTURES

$$f(x) = c^T x \quad \frac{\partial f}{\partial x} = \underline{c^T}_{\text{const.}}$$

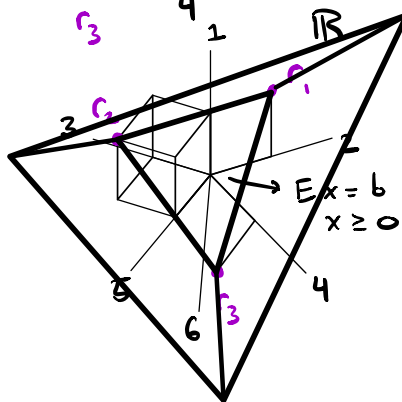
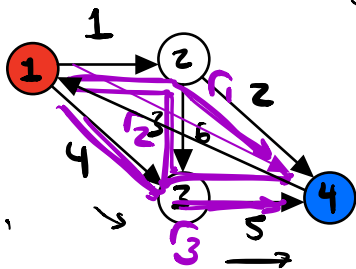
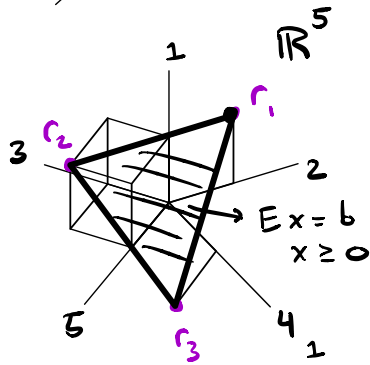
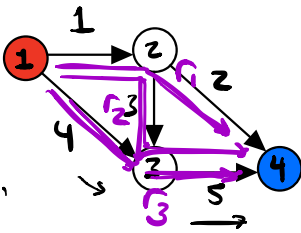
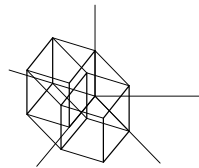
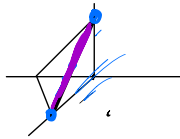
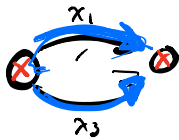
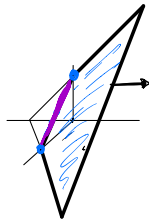
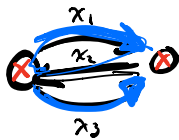
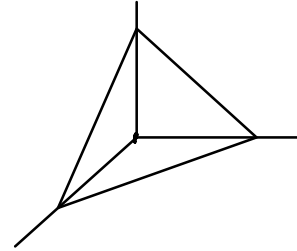
$$\min c^T x$$

$$\text{s.t. } \boxed{Ex = b} \quad \boxed{x \geq 0}$$

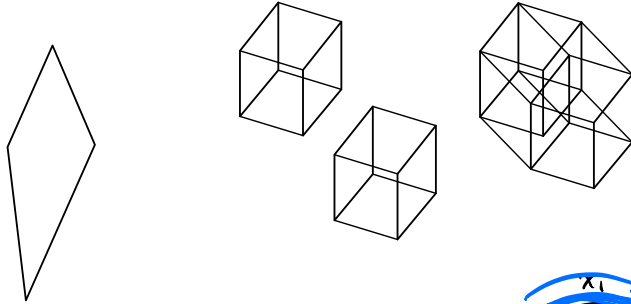


$$\begin{bmatrix} 1 & 1 & 1 \\ -1 & -1 & -1 \end{bmatrix} x = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

$$(1 \ 1 \ 1) x = 1$$



4 nodes:
complete graph w
4 nodes



$$\min_x C^T x$$

$$\text{s.t. } \underline{E} x = b \quad x \geq 0$$

$$C = \begin{bmatrix} c \\ -I \end{bmatrix}$$

$$\min_{x, z} C^T x$$

$$\text{s.t. } x = C z + d \quad x \geq 0$$

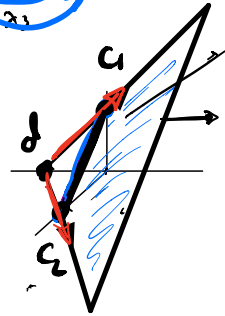
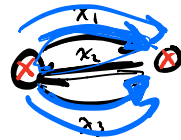
$$d: E d = b \quad C = [c_1 \ c_2]$$

list out the routes

$$R_1, R_2 \quad R = [R_1 \ R_2]$$

$$\left[\begin{array}{l} \min_{x, z} C^T x \\ \text{s.t. } x = R z \quad \mathbf{1}^T z = 1 \quad z \geq 0 \end{array} \right.$$

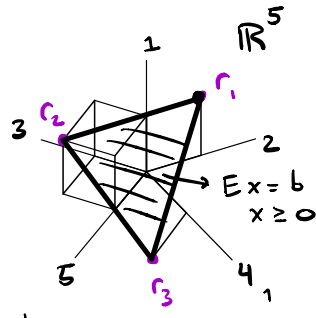
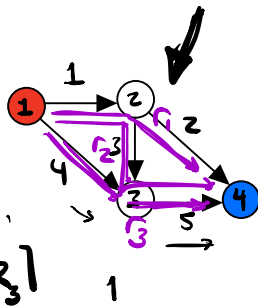
$$\left[\begin{array}{l} \min_z C^T R z \\ \text{s.t. } \mathbf{1}^T z = 1 \quad z \geq 0 \end{array} \right. \rightarrow \text{explicitly listed routes}$$



← Not equivalent



$$\begin{aligned} \min_z \quad & c^T R z \leftarrow \\ \text{s.t.} \quad & \mathbb{1}^T z = L, z \geq 0 \end{aligned}$$



$$c^T R = l^T = [cR_1 \ cR_2 \ cR_3]$$

$$l \in \mathbb{R}^{|R|}$$

l_R : cost of taking route R

$$R = [R_1 \ R_2 \ R_3]$$

R : set of routes

Dual of Edge Flow optimization

$$\begin{aligned} \min_x \quad & c^T x \leftarrow \\ \text{s.t.} \quad & Ex = b, x \geq 0 \end{aligned} \left. \right\} \leftarrow$$

$$L(x, v, \mu) = c^T x + v^T (Ex - b) - \mu^T x$$

$$\begin{aligned} \min_x \quad & c^T x \\ \text{s.t.} \quad & Ex = b, x \geq 0 \end{aligned} = \min_x \left(\max_{v, \mu \geq 0} \underbrace{c^T x + v^T (Ex - b) - \mu^T x}_{\text{func of } x} \right) \\ \geq \max_{v, \mu \geq 0} \left(\min_x \underbrace{c^T x + v^T (Ex - b) - \mu^T x}_{\text{func of } v, \mu} \right) = \max_{v, \mu \geq 0} \quad \quad \quad \text{s.t.} \quad \quad \quad *$$

$$\left(\min_x \quad c^T x + v^T (Ex - b) - \mu^T x \right) \Rightarrow \underbrace{c^T + v^T E - \mu^T}_{\text{dual objective}} = 0 \quad \quad \quad \text{dual constraint}$$

$$0 \quad (c^T + v^T E - \mu^T) x - \boxed{v^T b} - \mu^T x$$

Dual Problem:

$$\begin{aligned} \max \quad & -v^T b = v_0 - v_d \\ v, \mu \geq 0 \\ \text{s.t.} \quad & c^T + v^T E - \mu^T = 0 \quad \mu \geq 0 \end{aligned}$$

v : cost to go at a state
 μ : ineff. of edge

$$\begin{aligned} \max \quad & -v^T b = v_0 - v_d \quad \leftarrow \text{max difference in value func between origin \& dest.} \\ v, \mu \geq 0 \\ \text{s.t.} \quad & c^T + v^T E - \mu^T = 0 \quad \mu \geq 0 \\ & v^T E_i - v^T E_0 \end{aligned}$$

$$v^T E_0 = c^T + v^T E_i - \mu^T$$

More next week

for ea. edge



$$v_s = c_e + v_{s'} - \mu_e, \quad \mu_e \geq 0$$

$$\rightarrow v_s \leq c_e + v_{s'}$$

like * at every state

cost to go at state $s \leq$ immediate future edge cost + cost to go at the next state s'

like the Bellman eqn but forces v_s to be a lower bound on the cost to go.

if we explicitly list out the routes...

$$\begin{aligned} \min_z c^T R z &= \min_z \left(\max_{\lambda, u \geq 0} c^T R z - \lambda (\mathbb{1}^T z - 1) - u^T z \right) \\ \text{s.t. } \mathbb{1}^T z &= 1, z \geq 0 \\ \lambda &\in \mathbb{R}, u \in \mathbb{R}^{|R|} \end{aligned}$$

$$l^T = c^T R \geq \max_{\lambda, u \geq 0} \left(\min_z c^T R z - \lambda (\mathbb{1}^T z - 1) - u^T z \right)$$

↓

$$c^T R - \lambda \mathbb{1}^T - u^T = 0$$

Dual problem

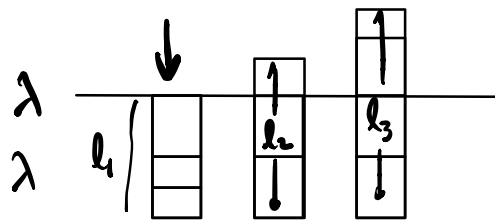
$$\max_{\lambda, u \geq 0} \lambda$$

$$\text{s.t. } \lambda \mathbb{1}^T + u^T = c^T R \quad u \geq 0$$

$$\left[\begin{array}{l} \max \lambda \\ \lambda \\ \text{s.t. } \lambda \mathbb{1}^T \leq c^T R = \underline{l}^T \end{array} \right. \quad \begin{array}{l} l = [l_1 \ l_2 \ l_3] \\ \text{cost for ea route} \end{array}$$

$$\lambda \leq l_r$$

Similar to the simplex setup we did last week



(R_1) R_2 R_3
↑ optimal route.

From last lecture
Simplex.

$$\begin{aligned} \rightarrow \max_z & \quad l^T z \\ \text{s.t.} & \quad \mathbb{1}^T z = 1, z \geq 0 \end{aligned}$$



dual

$$\begin{aligned} \rightarrow \min_{\lambda} & \quad \lambda \\ \text{s.t.} & \quad \lambda \mathbb{1}^T \geq l^T \end{aligned}$$

