

Two-stage Stochastic Programming

On average:

Minimum Spanning Trees

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Two stage

Stochastic Programming:

Input: situation today

distribution of situations tomorrow

Output: policy for today & all
possible tomorrows.

Objective: Minimize cost today
+ expected cost tomorrow

Example: 2-stage Matching

Input: costs on Monday: C_M

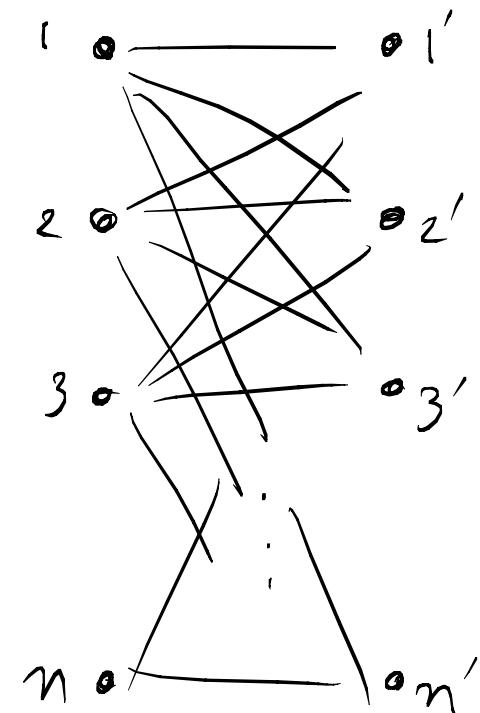
dist. of costs on Tuesday:

$$C_T(e) \sim \text{Exp}(1)$$

Output: Edges to buy on

Monday (because policy on Tuesday
is obvious — finish the matching.)

Soln: Find opt in poly time
by



Prior Work: (Lots)

<http://stoprog.org/>

R. Schultz, *Stochastic Integer Programming: A Tutorial*, 9th International Conference on Stochastic Programming, 2001.

M. H. van der Vlerk, *Stochastic Integer Programming Bibliography*, 1996-2003,
<http://mally.eco.rug.nl/stopbib.html>.

Combinatorial Problems:

M. Albareda-Sambola, M. H. van der Vlerk, E. Fernández, Exact solutions to a class of stochastic generalized assignment problem, Stochastic Programming E-Print Series, 2002.

A. Gupta, M. Pal, R. Ravi, A. Sinha, Boosted sampling: Approximation algorithms for stochastic optimization, STOC 2004.

N. Immorlica, D. Karger, M. Minkoff, V. Mirrokni, On the costs and benefits of procrastination: Approximation algorithms for stochastic combinatorial optimization problems, SODA 2004.

N. Kong, A. Schaefer, A factor $\frac{1}{2}$ approximation algorithm for two-stage stochastic matching problems, submitted.

R. Ravi, A. Sinha, Hedging uncertainty: Approximation algorithms for stochastic optimization problems, IPCO 2004.

Not the same model:

H. Ishii, S. Shiode, T. Nishida, Y. Namasuya, Stochastic spanning tree problem, *Discrete Appl. Math.* 3(4):263-273, 1981.

Minimum Spanning Trees:

Input: A graph $G = (V, E)$,
a cost vector $c \in \mathbb{R}^E$.

Output: A spanning tree $T \subseteq E$

Objective: $Z_{\text{opt}} = \text{minimize} \sum_{e \in T} c_e$

Many ways to find optimal value.

→ Often attributed to Kruskal (1956) and
Prim (1957), but dates to Boruvka (1926).

See R. Graham, P. Hell, "History of the min. sp. tree prob."

Random Minimum Spanning Trees:

If $c_e \sim U[0, 1]$ then

as $n \rightarrow \infty$,

$$Z_{\text{opt}} \rightarrow \zeta(3) = \frac{1}{1^3} + \frac{1}{2^3} + \frac{1}{3^3} + \dots \\ \approx 1.20205\dots$$

[Frieze, 1985]

Sketch of Proof:

length of tree T

$$\Rightarrow l(T) = \sum_{e \in T} x_e$$

$$= \sum_{e \in T} \int_0^1 \mathbb{1}_{\{x_e \geq p\}} dp$$

$$= \int_0^1 \sum_{e \in T} \mathbb{1}_{\{x_e \geq p\}} dp$$

for min. sp.
tree

$$= \int_0^1 (k(G_p) - 1) dp \quad \text{\# of components}$$

$$E[l(T)] = \int_{p=0}^1 E[k(G_p)] dp - 1$$

only trees/contribute
and giants

$$\begin{aligned} & \approx \int_{p=0}^1 \sum_{k=0}^n \binom{n}{k} k^{k-2} p^{k-1} (1-p)^{kn} dp \\ & \approx \int_{p=0}^1 \sum_{k=0}^n \frac{n^k}{k!} k^{k-2} p^{k-1} (1-p)^{kn} dp \\ & \approx \sum_{k=0}^n \frac{n^k}{k!} k^{k-2} \left[\int_0^1 p^{k-1} (1-p)^{kn} dp \right] \\ & \quad \left(\frac{(k-1)! (kn)!}{(k(n+1))!} \right) \quad \text{Beta Integral} \\ & \approx \sum_k k^{-3} \end{aligned}$$

2-Stage Stochastic Minimum Spanning Trees:

Input: A cost vector C_M
A distribution over cost vectors C_T

Output: A forest F to buy Monday
(since best policy on Tuesday is obvious)

Objective: $Z_{opt} = \text{minimize } \sum C_M(e)$
 $+ E \left[\min_{F'} \left\{ \sum_{e \in F'} C_T(e) : F \cup F' \text{ is a spanning tree} \right\} \right]$

NP-hard, hard to approx. beyond $O(\log n)$.

Random 2-Stage Stochastic Min. Sp. Trees:

$$c_M(e) \sim U[0, 1]$$

$$c_T(e) \sim U[0, 1]$$

Same question.

Some observations :

① Buying a spanning tree all on Monday costs $\ell(3)$.

② If you knew Tuesday's costs on Monday, could get $\ell(3)/2$.

Random 2-Stage Stochastic Min. Sp. Trees:

A reasonable approach:

Pick some threshold value, α .

- On Monday, only buy edges with cost less than α .
- On Tuesday, finish the tree.

Best value $\alpha = \frac{1}{n}$

Objective value $Z_{\text{threshold } \frac{1}{n}} = \zeta(3) - \frac{1}{2}$.

Random 2-Stage Stochastic Min. Sp. Trees:

- Threshold is not optimal:

By looking at structure instead of just cost, can improve a little.

$$z_{\text{opt}} \leq \zeta(3) - \frac{1}{2} - 10^{-256}.$$

- There is no way to attain $\zeta(3)/2$, because you must make some mistakes on Monday.

$$z_{\text{opt}} \geq \zeta(3)/2 + 10^{-5}.$$

Random 2-Stage Stochastic Min. Sp. Trees:

Rooted Arborescence
(directed version of sp. tree)

Threshold heuristic now is:

- On Monday, for all v , if cheapest out-edge costs less than α and does not form a cycle, buy it.
- On Tuesday, complete Rooted Arbor. as cheaply as possible.

Best $\alpha = \frac{1}{n-1}$, which makes $Z_{\text{opt}} = 1 - \frac{1}{e}$

Random 2-Stage Stochastic Min. Rooted Arb:

- For rooted Arborescence, threshold heuristic is optimal.

Pf: Every vertex ~~has~~ ^{besides the root} to have an out-edge.

Just look at a single vertex.

The expected cost of the min. out-edge on Tuesday is $\frac{1}{n}$.

So the ~~cost~~ ^{expected} of the out-edges is at least

$$(n-1) E \left[\min \left\{ \frac{1}{n}, c_M(e_1), c_M(e_2), \dots, c_M(e_{n-1}) \right\} \right]$$
$$\approx 1 - \frac{1}{e}.$$

One more thing:

Concentration: Say $Z = \text{cost of MST on}$
rand. $[0, 1]$
edge wts.

Want a bound like

$$\Pr[|Z - \zeta(3)| \geq \lambda] \leq e^{-\delta n}.$$

Where (since $E[Z] = \zeta(3)$ is constant)
we have λ & δ constant as well.

We prove this using a great inequality:

Symmetrized Logarithmic Sobolev Ineq.:

$Z = f(X_1, \dots, X_N)$ where all X_i i.i.d.

$Z'_i = f(X_1, \dots, \overset{\text{copy of } X_i}{X'_i}, \dots, X_N)$ where X'_i is indep

If $E\left[\sum_{i=1}^N (Z - Z'_i)^2 \mathbb{1}_{Z > Z'_i} | X_1, \dots, X_N\right] \leq c$,

then for all $t > 0$,

$$\Pr[Z > E[Z] + t] \leq e^{-t^2/4c}.$$