

Robust Optimization with Recovery: Application to Shortest Paths and Airline Scheduling

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Planning the Recovery

- Developed Recovery Algorithm for disrupted airline schedules
- Want to consider the possibility of recovery during scheduling phase

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- Proactive worst case approach including recovery decisions
- Illustration on the SPPID
- Application to Airline Scheduling
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Optimization under Uncertainty

(for a minimization problem)

- Characterization of **uncertainty set U**
 - any knowledge?
 - probabilistic measure?
 - consider extreme cases?

- **Proactive** vs **Reactive** approach

Existing Methods' Classification

	Proactive Methods	Reactive Methods
Probabilistic U	Stochastic Programming	Stochastic Programming with Recourse
Non Probabilistic U	Robust Optimization	On-line Algorithms

Some References

- On-Line Algorithms: **Albers**, 2003
- Stochastic Optimization: **Kall and Wallace**, 1994; **Wallace and Ziemba**, 2005.
- Stochastic with Recourse: **Kall and Wallace**, 1994; **Polychronopoulos and Tsitsiklis** , 1996; **Provan**, 2003.
- Robust Optimization: **Soyster**, 1973; **Ben-Tal and Nemirovski**, 2001 (book); **Bertsimas and Sim**, 2004.

On-Line Optimization

Advantages

- react in real time
- easy computation (decision strategy)
- optimality ratio (a posteriori)

Disadvantages

- no information on bounds (only ratio)
- reaction dictated by nature
- difficult to measure performance a priori

Stochastic Programming

Advantages	Disadvantages
<ul style="list-style-type: none">• lowest expected cost• low probabilities of high cost• exploits a priori information	<ul style="list-style-type: none">• difficult to measure probabilities• estimation of all scenarios• only predictor (depends on scenario)• not efficient for few implementations

Stochastic Programming with Recourse

Advantages	Disadvantages
<ul style="list-style-type: none">• includes cost of recourse decision• adapts to information in real time• exploits partial information	<ul style="list-style-type: none">• difficult to measure probabilities• estimation of all scenarios• predictor on partially revealed scenario• guided by information revelation

Robust Optimization

Advantages	Disadvantages
<ul style="list-style-type: none">• solution always feasible• evaluation on worst scenarios only• get cost bound	<ul style="list-style-type: none">• characterize worst scenarios• what if a good scenario occurs?

Definitions

Problem P

- set of **feasible solutions** S
- uncertainty set U
- **realization** or **scenario** $u \in U$
- cost of solution s under scenario u : $c_u(s)$

Recoverable Approach

- **Proactive**: want a deterministic solution not dictated by scenario revealing
- **Worst + Best**: consider both best and worst scenarios (**potential**)
- **Include On-line Decisions (Recovery)**: consider recovery costs for unfeasible solutions

$$(P') \quad \min_{s \in S} \left\{ \max_{u \in U} \tilde{c}_u(s) + c_u^{\text{REC}}(s) + \min_{u' \in U} \tilde{c}_{u'}(s) + c_{u'}^{\text{REC}}(s) \right\}$$

Example:

Shortest Path Problem with Interval Data (SPPID):

- Oriented graph $G = (V, A)$
- Unique **source** s and unique **sink** t
- **Random** arc costs $c_{ij} \in [l_{ij}, u_{ij}]$ ($u_{ij} = \infty$ possible)
- Scenario: set of cost realization for every arc cost
- arc cost revealed when source node is reached

On-Line Strategy:

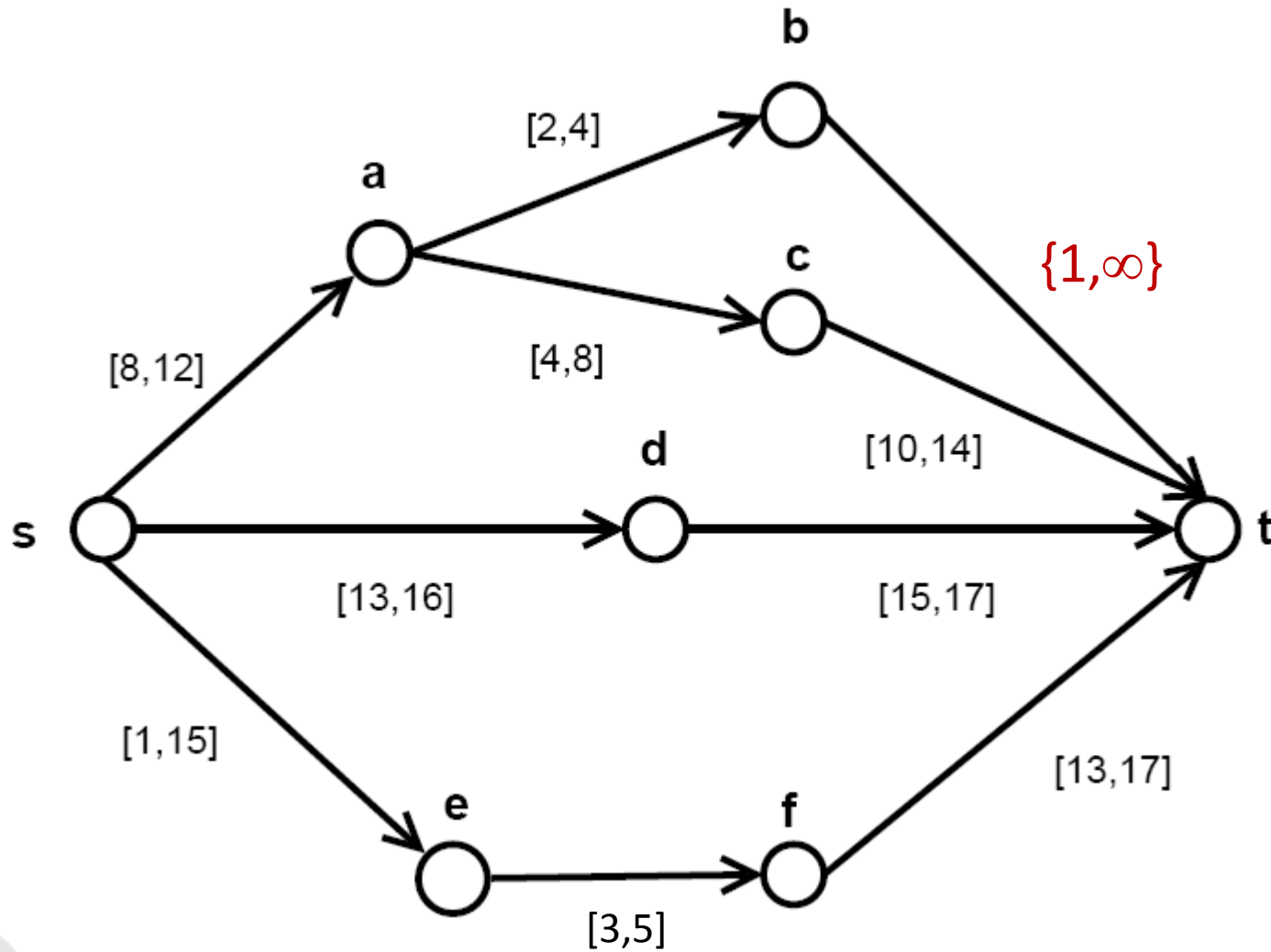
- at node i take outgoing arc with lowest cost

Recovery:

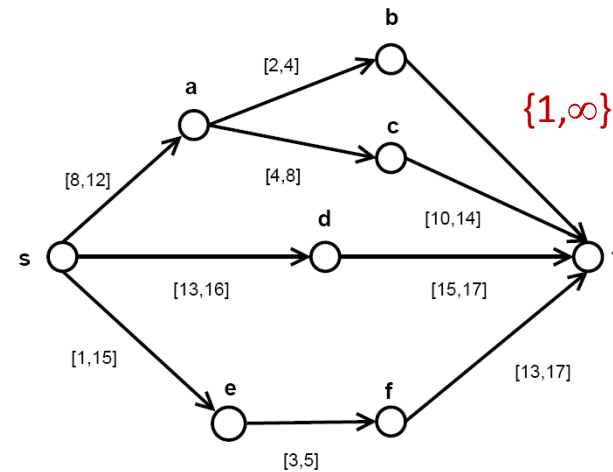
- when reach dead end, take last arc **backwards** at cost u_{ij}
- **remove arc** from V

Probabilistic cases:

- symmetric independent distributions

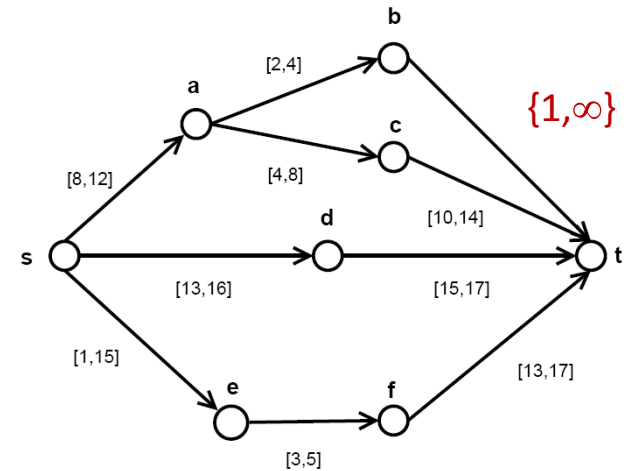


- **Stochastic**: path $\{s, e, f, t\}$, mean cost 27
- **Robust**: path $\{s, d, t\}$, worst cost 33
- **On-line**: either $\{s, a, b, t\}$ or $\{s, e, f, t\}$
- **Stochastic with recourse**: either $\{s, e, f, t\}$ or $\{s, a, b, t\}$

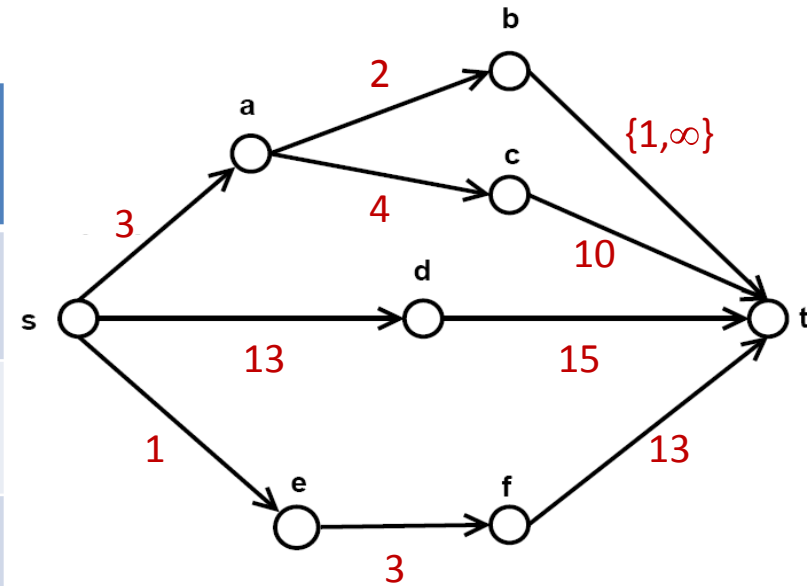


- **Recoverable:** path $\{s, a, b, t\}$

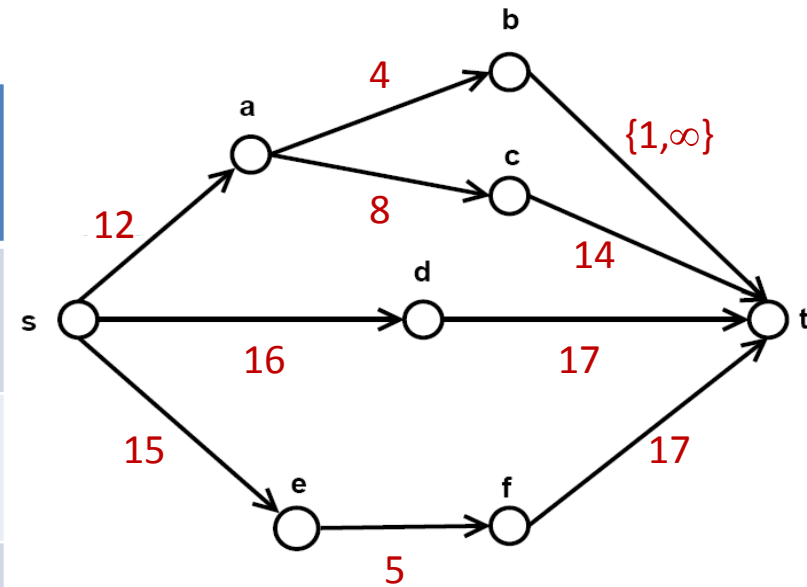
Path	Best	Worst	Potential	Mean
$\{s, a, b, t\}$	11	42	53	∞
$\{s, a, c, t\}$	22	34	56	28
$\{s, d, t\}$	28	33	61	30.5
$\{s, e, f, t\}$	17	37	54	27



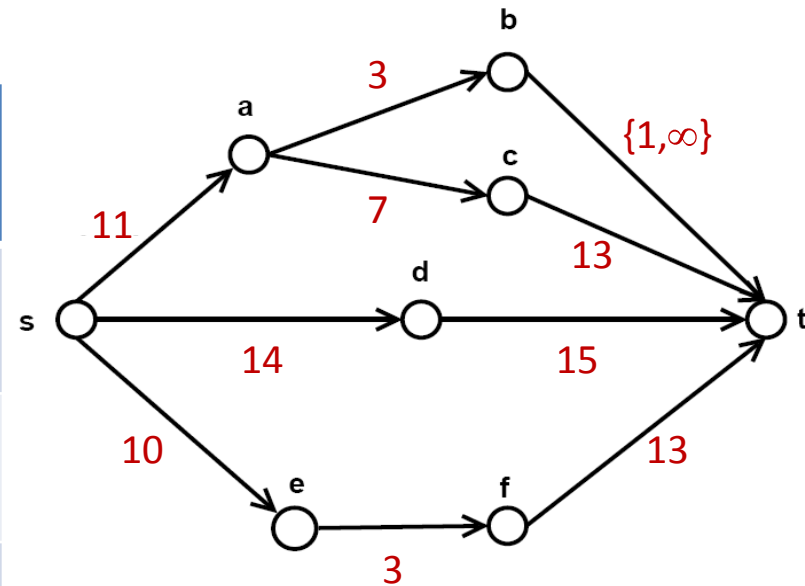
Method	Path	Cost
On-Line	$\{s, e, f, t\}$	17
Stochastic with Recourse	$\{s, e, f, t\}$	17
Stochastic	$\{s, e, f, t\}$	17
Robust	$\{s, d, t\}$	28
Recoverable	$\{s, a, b, t\}$	11
	Or $\{s, a, b, a, c, t\}$	or 28



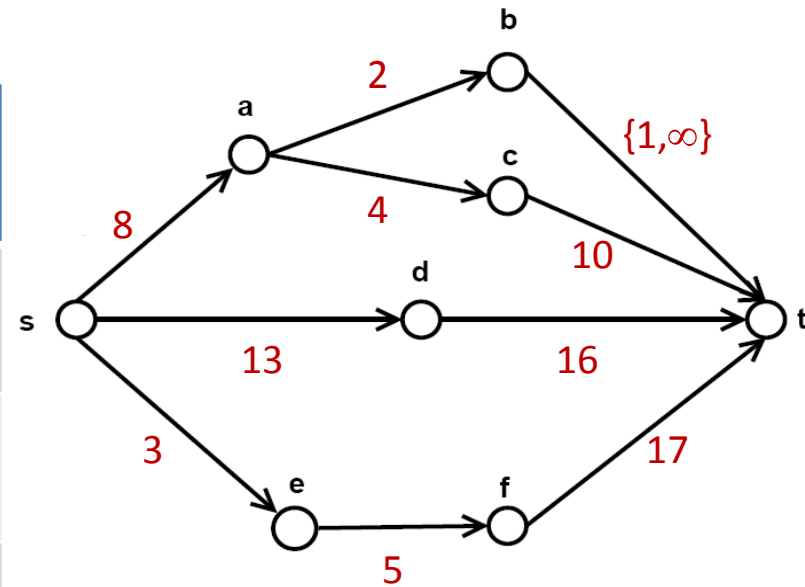
Method	Path	Cost
On-Line	$\{s, a, b, t\}$	17
	$\{s, a, b, a, c, t\}$	42
Stochastic with Recourse	$\{s, a, b, t\}$	17
	$\{s, a, b, a, c, t\}$	42
Stochastic	$\{s, e, f, t\}$	37
Robust	$\{s, d, t\}$	33
Recoverable	$\{s, a, b, t\}$	17
	$\{s, a, b, a, c, t\}$	42



Method	Path	Cost
On-Line	$\{s, e, f, t\}$	26
Stochastic with Recourse	$\{s, a, b, t\}$ or $\{s, a, b, a, c, t\}$	15 38
Stochastic	$\{s, e, f, t\}$	26
Robust	$\{s, d, t\}$	29
Recoverable	$\{s, a, b, t\}$ or $\{s, a, b, a, c, t\}$	15 38



Method	Path	Cost
On-Line	$\{s, e, f, t\}$	25
Stochastic with Recourse	$\{s, e, f, t\}$	25
Stochastic	$\{s, e, f, t\}$	25
Robust	$\{s, d, t\}$	29
Recoverable	$\{s, a, b, t\}$	11
	$\{s, a, b, a, c, t\}$	28



Average costs over 8 different scenarios

Method	Path	Cost
On-Line	Scenario dependent	26.875
Stochastic with Recourse	Scenario dependent	26.375
Stochastic	$\{s,e,f,t\}$	26.250
Robust	$\{s,d,t\}$	29.875
Recoverable	$\{s,a,b,t\}$ or $\{s,a,b,a,c,t\}$	24.250

- Recoverable solution both highest and lowest cost
- Robust and Stochastic reject possibly unfeasible paths
- Reactive dependent on scenario

REMARK

Path $\{s, a, b, t\}$ has least expected cost (24.5) if consider **stochastic with recourse** in a **proactive** way

Recoverable Schedule for Airlines

Difficulties:

- evaluation of feasibility
- recovery problem *NP*-hard
- characterization of worst scenario impossible
- evaluation on whole U impossible

Research Directions:

- Scenario Sampling
- Uncertainty set structure
- Recovery costs estimations
- Implicit measures and multi-objective optimization
- Explicit measures in recovery algorithm



Future Work:

- Develop a specific formalism
- Explore more deeply possible directions
- Develop an efficient algorithm
- Benchmark with real problems

Conclusions:

- **Proactive best-worst** case framework including information on unfeasible solutions
- Few scenario estimations, no probabilistic measures
- Provides recovery plan for each scenario
- Trade-off between proactive and reactive (**planning the recovery**)

THANKS for your attention!

Any Questions?

