

1.206J/16.77J/ESD.215J
Airline Schedule Planning

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1.206J/16.77J/ESD.215J Airline Schedule Planning: Multi-commodity Flows

Outline

- Applications
- Problem Definition
- Formulations
- Solutions
- Results

Application I

- Package flow problem (express package delivery operation)
 - Shipments have specific origins and destinations and must be routed over a transportation network
 - Each set of packages with a common origin-destination pair is called a commodity
 - Time windows (availability and delivery time) associated with packages
 - The objective might be to minimize total costs, find a feasible flow, ...

Application II

- Passenger mix problem
 - Given a fixed schedule of flights, a fixed fleet assignment and a set of customer demands for air travel service on this fleted schedule, the airline's objective is to maximize revenues by accommodating as many high fare passengers as possible
 - For some flights, demand exceeds seat supply and passengers must be *spilled* to other itineraries of either the same or another airline

Application III

- Message routing problem
 - In a telecommunications or computer network, requirements exist for transmission lines and message requests, or commodities.
 - The problem is to route the messages from their origins to their respective destinations at minimum cost

MCF Networks

- Set of nodes
 - Each node associated with the supply of or demand for commodities
- Set of arcs
 - Cost per unit commodity flow
 - Capacity limiting the *total* flow of all commodities and/ or the flow of individual commodities

MCF Commodity Definitions

- A commodity may originate at a subset of nodes in the network and be destined for another subset of nodes
- A commodity may originate at a single node and be destined for a subset of the nodes
- A commodity may originate at a single node and be destined for a single node

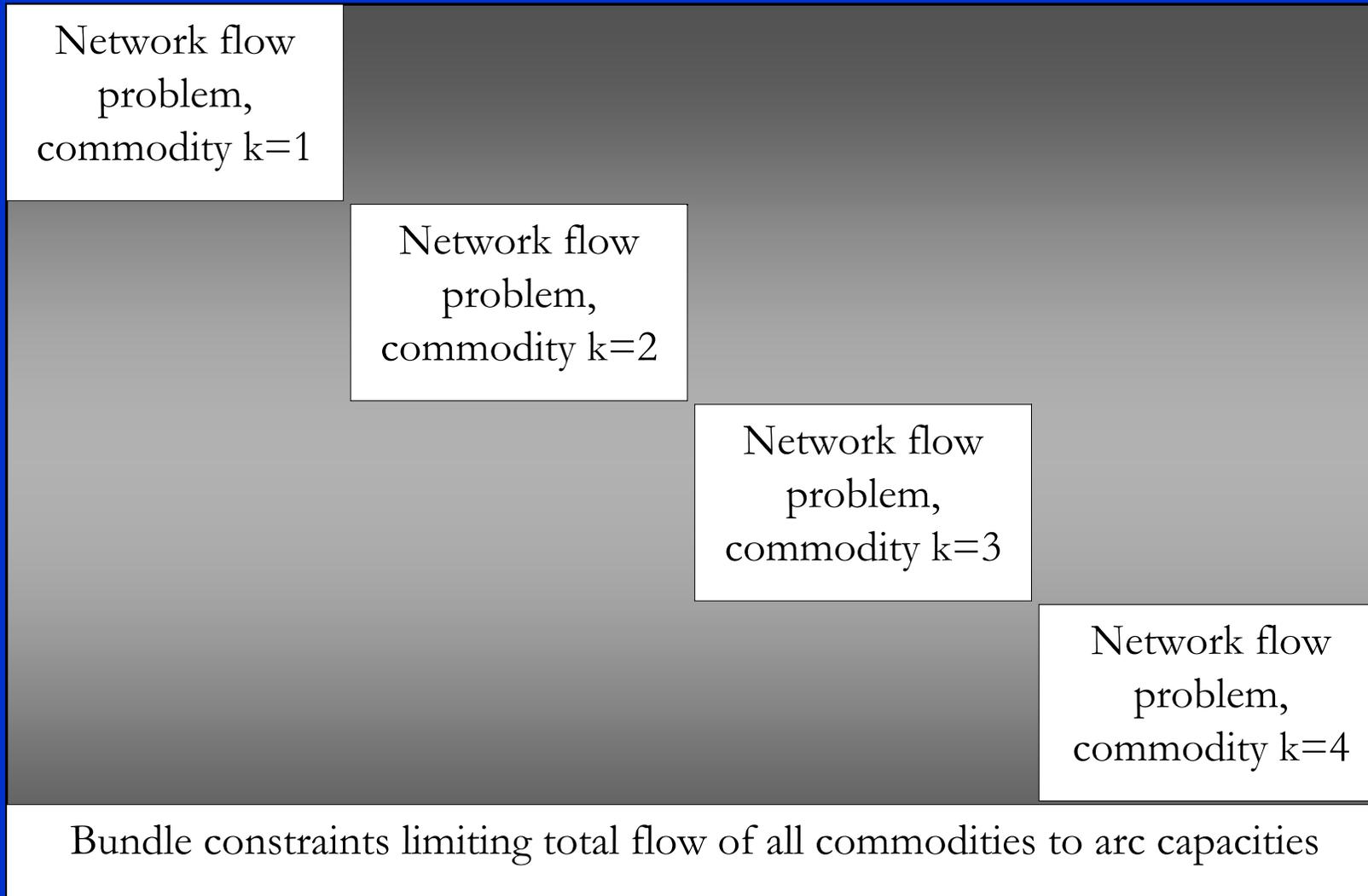
MCF Objectives

- Flow the commodities through the networks from their respective origins to their respective destinations at minimum cost
 - Expressed as distance, money, time, etc.
- Ahuja, Magnanti and Orlin (1993)-- survey of multi-commodity flow models and solution procedures

MCF Problem Formulations -- Linear Programs

- Network flow problems
 - Capacity constraints limit flow of individual commodities
 - Conservation of flow constraints ensure flow balance for individual commodities
 - Flow non-negativity constraints
- With side constraints
 - Bundle constraints restrict total flow of *ALL* commodities on an arc

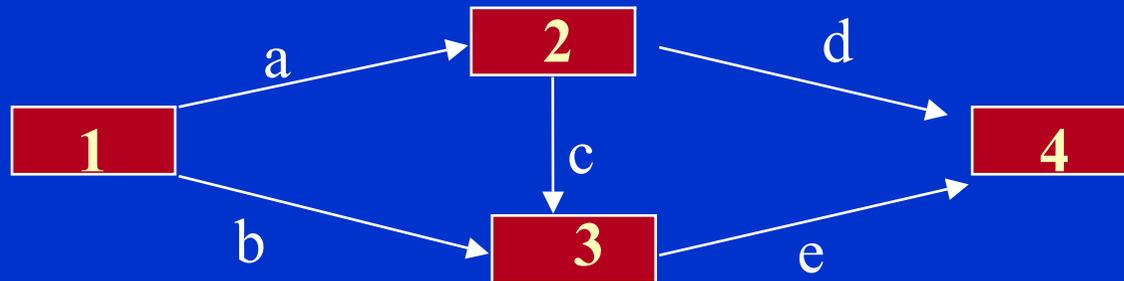
MCF Constraint Matrix



Alternative Formulations for O-D Commodity Case

- Node-Arc Formulation
 - Decision variables: flow of commodity k on each arc ij
- Path Formulation
 - Decision variables: flow of commodity k on each path for k
- “Tree” or “Sub-network” Formulation
 - Define: super commodity: set of all (O-D) commodities with the same origin o (or destination d)
 - Decision variables: quantity of the super commodity k' assigned to each “tree” or “sub-network” for k'
- Formulations are equivalent

Sample Network



Arcs

<u>i</u>	<u>j</u>	<u>cost</u>	<u>capy</u>
1	2	1	20
1	3	2	10
2	3	3	20
2	4	4	10
3	4	5	40

Commodities

<u>#</u>	<u>o</u>	<u>d</u>	<u>quant</u>
1	1	3	5
2	1	4	15
3	2	4	5
4	3	4	10

Notation

Parameters

- A : set of all network arcs
- K : set of all commodities
- N : set of all network nodes
- $O(k)$ [$D(k)$]: origin [destination] node for commodity k
- c_{ij}^k : per unit cost of commodity k on arc ij
- u_{ij} : total capacity on arc ij
(assume u_{ij}^k is unlimited for each k and each ij)
- d_k : total quantity of commodity k

Decision Variables

- x_{ij}^k : number of units of commodity k assigned to arc ij

Node-Arc Formulation

Minimize $\sum_{ij} \sum_k c_{ij}^k x_{ij}^k$

subject to

$$\sum_j x_{ij}^k - \sum_j x_{ji}^k = d_k \quad \text{if } i \in O(k)$$

$$= -d_k \quad \text{if } i \in D(k)$$

$$= 0 \quad \text{otherwise}$$

: Conservation of Flow

$$\sum_k x_{ij}^k \leq u_{ij} \quad \forall (i, j) \in A$$

: Bundle constraints

$$x_{ij}^k \geq 0 \quad \forall (i, j) \in A, k \in K$$

: Nonnegativity constraints

	k 1					k 2					k 3					k 4					RHS
	a	b	c	d	e	a	b	c	d	e	a	b	c	d	e	a	b	c	d	e	
1	1	1																			= d ₁
2	-1																				= 0
3		-1	-1																		= -d ₁
4				-1	-1																= 0
1						1	1														= d ₂
2						-1															= 0
3							-1	-1													= 0
4									-1	-1											= -d ₂
1											1	1									= 0
2											-1										= d ₃
3												-1	-1								= 0
4														-1	-1						= -d ₃
1																1	1				= 0
2																-1					= 0
3																	-1	-1			= d ₄
4																			-1	-1	= -d ₄
a	1					1					1					1					≤ u _a
b		1					1					1					1				≤ u _b
c			1					1					1					1			≤ u _c
d				1					1					1					1		≤ u _d
e					1					1					1					1	≤ u _e
	c _a ¹ x _a ¹	c _b ¹ x _b ¹	c _c ¹ x _c ¹	c _d ¹ x _d ¹	c _e ¹ x _e ¹	c _a ² x _a ²	c _b ² x _b ²	c _c ² x _c ²	c _d ² x _d ²	c _e ² x _e ²	c _a ³ x _a ³	c _b ³ x _b ³	c _c ³ x _c ³	c _d ³ x _d ³	c _e ³ x _e ³	c _a ⁴ x _a ⁴	c _b ⁴ x _b ⁴	c _c ⁴ x _c ⁴	c _d ⁴ x _d ⁴	c _e ⁴ x _e ⁴	

Additional Notation

Parameters

- P^k : set of all paths for commodity k , for all k
- c_p : per unit cost of commodity k on path p
 $= \sum_{ij \in p} c_{ij}^k$
- δ_{ij}^p : = 1 if path p contains arc ij ; and = 0 otherwise

Decision Variables

- f_p : fraction of total quantity of commodity k assigned to path p

O/D Based Path Formulation

Minimize
$$\sum_k \sum_{p \in P^k} d_k C_p f_p$$

subject to

$$\sum_k \sum_{p \in P^k} d_k f_p \delta_{ij}^p \leq u_{ij} \quad \forall (i, j) \in A \quad \text{: Bundle constraints}$$

$$\sum_{p \in P^k} f_p = 1 \quad \forall k \in K \quad \text{: Flow balance constraints}$$

$$f_p \geq 0 \quad \forall p \in P^k, k \in K \quad \text{: Non-neg. constraints}$$

	Path								RHS	Dual
	k=1		k=2			k=3		k=4		
a	d ₁	0	d ₂	d ₂	0	0	0	0	≤ u _a	-π _a
b	0	d ₁	0	0	d ₂	0	0	0	≤ u _b	-π _b
c	d ₁	0	d ₂	0	0	d ₃	0	0	≤ u _c	-π _c
d	0	0	0	d ₂	0	0	d ₃	0	≤ u _d	-π _d
e	0	0	d ₂	0	d ₂	d ₃	0	d ₄	≤ u _e	-π _e
k=1	1	1							= 1	σ ¹
k=2			1	1	1				= 1	σ ²
k=3						1	1		= 1	σ ³
k=4								1	= 1	σ ⁴
Cost.	C ₁ d ₁	C ₂ d ₁	C ₃ d ₂	C ₄ d ₂	C ₅ d ₂	C ₆ d ₃	C ₇ d ₃	C ₈ d ₃		
Variable	f ₁	f ₂	f ₃	f ₄	f ₅	f ₆	f ₇	f ₈		

Additional Notation

Parameters

- S : set of source nodes
 $n \in N$ for all commodities
- Q^s : the set of all sub-networks originating at s
- TC_q^s : total cost of sub-network q originating at s
- ζ_p^q : = 1 if path p is contained in sub-network q ; and = 0 otherwise

Decision Variables

- R_q^s : fraction of total quantity of the super commodity originating at s assigned to sub-network q

Sub-network Formulation

$$\text{Minimize } \sum_{s \in S} \sum_{q \in Q^s} \left(\sum_{k \in q} \sum_{p \in P^k} c_p \zeta_p^q d_k \right) R_q^s$$

subject to

$$\sum_s \sum_{q \in Q^s} \left(\sum_{k \in s} \sum_{p \in P^k} d_k \delta_{ij}^p \right) R_q^s \zeta_p^q \leq u_{ij} \quad \forall (i, j) \in A \quad \text{: Capacity Limits on Each Arc}$$

$$\sum_{q \in Q^s} R_q^s = 1 \quad \forall s \in S \quad \text{: Mass Balance Requirements}$$

$$R_q^s \geq 0 \quad \forall q \in Q^s, s \in S \quad \text{: Nonnegative Path Flow Variables}$$

	Sub- network									RHS	Dual
	o=1			o=2			o=3				
a	d ₁ +d ₂	d ₁ +d ₂	d ₁	d ₂	d ₂	0	0	0	0	<= u _a	π _a
b	0	0	d ₂	d ₁	d ₁	d ₁ +d ₂	0	0	0	<= u _b	π _b
c	d ₁	d ₁ +d ₂	d ₁	0	d ₂	0	d ₃	0	0	<= u _c	π _c
d	d ₂	0	0	d ₂	0	0	0	d ₃	0	<= u _d	π _d
e	0	d ₂	d ₂	0	d ₂	d ₂	d ₃	0	d ₄	<= u _e	π _e
o=1	1	1	1	1	1	1				= 1	σ ¹
o=2							1	1		= 1	σ ²
o=3									1	= 1	σ ³
Cost.	TC ₁ ¹	TC ₂ ¹	TC ₃ ¹	TC ₄ ¹	TC ₅ ¹	TC ₆ ¹	TC ₁ ²	TC ₂ ²	TC ₁ ³		
Variable	R ₁ ¹	R ₂ ¹	R ₃ ¹	R ₄ ¹	R ₅ ¹	R ₆ ¹	R ₁ ²	R ₂ ²	R ₁ ³		

Linear MCF Problem Solution

- Obvious Solution
 - LP Solver
- Difficulty
 - Problem Size: ($|N| = |\text{Nodes}|$, $|C| = |\text{Commodities}|$, $|A| = |\text{Arcs}|$)
 - Node-arc formulation:
 - Constraints: $|N| * |C| + |A|$
 - Variables: $|A| * |C|$
 - Path formulation:
 - Constraints: $|A| + |C|$
 - Variables: $|\text{Paths for ALL commodities}|$
 - Sub-network formulation:
 - Constraints: $|A| + |\text{Origins}|$
 - Variables: $|\text{Combinations of Paths by Origin}|$

General MCF Solution Strategy

- Try to Decompose a Hard Problem Into a Set of Easy Problems

MCF Solution Procedures I

- Partitioning Methods
 - Exploit Network Structure to Speed Up Simplex Matrix Computations
- Resource-Directive Decomposition
 - Repeat until Optimal:
 - Allocate Arc Capacity Among Commodities
 - Find Optimal Flows Given Allocation

MCF Solution Procedures II

- Price-Directive Decomposition
 - Repeat until Optimal:
 - Modify Flow Cost on Arc
 - Ignore Bundle Constraints, Find Optimal Flows

Revisiting the Path Formulation

$$\text{MINIMIZE } \sum_{k \in K} \sum_{p \in P^k} d_k c_p f_p$$

$$\text{subject to: } \sum_{p \in P^k} \sum_{k \in K} d_k f_p \delta_{ij}^p \leq u_{ij} \quad \forall ij \in A$$

$$\sum_{p \in P(k)} f_p = 1 \quad \forall k \in K$$

$$f_p \geq 0 \quad \forall p \in P^k, \forall k \in K$$

By-products of the Simplex Algorithm: Dual Variable Values

Duals

$-\pi_{ij}$: the dual variable associated with the bundle constraint for arc ij (π is non-negative)

σ^k : the dual variable associated with the commodity constraints

Economic Interpretation

π_{ij} : the value of an additional unit of capacity on arc ij

σ^k / d_k : the minimal cost to send an additional unit of commodity k through the network

Modified Costs

Definition: Modified cost for arc ij and commodity $k = c_{ij}^k + \pi_{ij}$

Definition: Modified cost for path p and commodity $k = \sum_{ij \in A} (c_{ij}^k + \pi_{ij}) \delta_{ij}^p$

Optimality Conditions for the Path Formulation

f_p^* and π_{ij}^* , σ^{*k} are optimal for all k and all ij iff:

Primal feasibility is satisfied

1. $\sum_{p \in P^k} \sum_{k \in K} d_k f_p^* \delta_{ij}^p \leq u_{ij} \quad \forall ij \in A$
2. $\sum_{p \in P(k)} f_p^* = 1 \quad \forall k \in K$
3. $f_p^* \geq 0 \quad \forall p \in P^k, \forall k \in K$

Complementary slackness is satisfied

1. $\pi_{ij}^* (\sum_{p \in P^k} \sum_{k \in K} d_k f_p^* \delta_{ij}^p - u_{ij}) = 0, \quad \forall ij \in A$
2. $\sigma^{*k} (\sum_{p \in P^k} f_p^* - 1) = 0, \quad \forall k \in K$

Dual feasibility is satisfied (reduced cost is non-negative for a minimization problem)

1. $(d_k c_p + \sum_{ij \in A} d_k \pi_{ij} \delta_{ij}^p) - \sigma^k = d_k (\sum_{ij \in A} (c_{ij}^k + \pi_{ij}) \delta_{ij}^p - \sigma^k / d_k) \geq 0, \quad \forall p \in P^k, \forall k \in K$

Multi-commodity Flow Optimality Conditions

- The price for an additional unit of capacity is 0 unless capacity is fully utilized

$$1. \pi^*_{ij} (\sum_{p \in P^k} \sum_{k \in K} d_k f^*_p \delta^p_{ij} - u_{ij}) = 0, \quad \forall ij \in A$$

- A path p for commodity k is utilized only if its “modified cost” (that is, $\sum_{ij \in A} (c_{ij}^k + \pi^*_{ij} \delta^p_{ij})$) is minimal, for all paths $p \in P^k$

1. *Reduced Costs all non-negative:*

$$c'_p = d_k (\sum_{ij \in A} (c_{ij}^k + \pi^*_{ij}) \delta^p_{ij} - \sigma^{*k} / d_k) \geq 0, \\ \forall p \in P^k, \forall k \in K$$

$$2. f^*_p (\sum_{ij \in A} (c_{ij}^k + \pi^*_{ij}) \delta^p_{ij} - \sigma^{*k} / d_k) = 0, \\ \forall p \in P^k, \forall k \in K$$

Column Generation- A Price Directive Decomposition

Millions/Billions of Variables



RMP and Optimality Conditions

Consider f_p^* and π_{ij}^* , σ^{*k} optimal for RMP, then

Primal feasibility is satisfied

1. $\sum_{p \in P^k} \sum_{ij \in A} d_k f_p^* \delta_{ij}^p \leq u_{ij} \quad \forall ij \in A$
2. $\sum_{p \in P(k)} f_p^* = 1 \quad \forall k \in K$
3. $f_p^* \geq 0 \quad \forall p \in P^k, \forall k \in K$

Complementary slackness is satisfied

1. $\pi_{ij}^* (\sum_{p \in P^k} \sum_{ij \in A} d_k f_p^* \delta_{ij}^p - u_{ij}) = 0, \quad \forall ij \in A$
2. $\sigma^{*k} (\sum_{p \in P^k} f_p^* - 1) = 0, \quad \forall k \in K$

Dual feasibility is guaranteed (reduced cost is non-negative) ONLY for a path p included in RMP

1. $(d_k c_p + \sum_{ij \in A} d_k \pi_{ij}^* \delta_{ij}^p) - \sigma^{*k} = d_k (\sum_{ij \in A} (c_{ij}^k + \pi_{ij}^*) \delta_{ij}^p - \sigma^{*k} / d_k) \geq 0, \quad \forall p \in P^k, \forall k \in K$

LP Solution: Column Generation

- Step 1: Solve *Restricted Master Problem* (RMP) with subset of all variables (columns)
- Step 2: Solve *Pricing Problem* to determine if any variables when added to the RMP can improve the objective function value (that is, if any variables have negative reduced cost)
- Step 3: If variables are identified in Step 2, add them to the RMP and return to Step 1; otherwise STOP

Pricing Problem

- Given π , the optimal (non-negative) duals for the current restricted master problem, the pricing problem, for each $p \in P^k$, $k \in K$ is

$$\min_{p \in P^k} (d_k (\sum_{ij \in A} (c_{ij}^k + \pi_{ij}) \delta_{ij}^p - \sigma^k / d_k))$$

Or, equivalently:

$$\min_{p \in P^k} \sum_{ij \in A} (c_{ij}^k + \pi_{ij}) \delta_{ij}^p$$

➤ *A shortest path problem for commodity k (with modified arc costs)*

Example- Iteration 1

	Path								RHS	Dual
	k=1		k=2			k=3		k=4		
a	5	0	15	15	0	0	0	0	≤ 20	$\pi_a = 0$
b	0	5	0	0	15	0	0	0	≤ 10	$\pi_b = 0$
c	5	0	15	0	0	5	0	0	≤ 20	$\pi_c = 0$
d	0	0	0	15	0	0	5	0	≤ 10	$\pi_d = 0$
e	0	0	15	0	15	5	0	10	≤ 40	$\pi_e = 0$
k=1	1	1							= 1	$\sigma^1 = 10$
k=2			1	1	1				= 1	$\sigma^2 = 135$
k=3						1	1		= 1	$\sigma^3 = 20$
k=4								1	= 1	$\sigma^4 = 50$
Cost.	20	10	135	75	105	40	20	50		
Variable	f_1	$f_2 = 1$	$f_3 = 1$	f_4	f_5	f_6	$f_7 = 1$	$f_8 = 1$		

Example- Iteration 2

	Path								RHS	Dual
	k=1		k=2			k=3		k=4		
a	5	0	15	15	0	0	0	0	≤ 20	$\pi_a = 0$
b	0	5	0	0	15	0	0	0	≤ 10	$\pi_b = 2$
c	5	0	15	0	0	5	0	0	≤ 20	$\pi_c = 0$
d	0	0	0	15	0	0	5	0	≤ 10	$\pi_d = 4$
e	0	0	15	0	15	5	0	10	≤ 40	$\pi_e = 0$
k=1	1	1							= 1	$\sigma^1 = 20$
k=2			1	1	1				= 1	$\sigma^2 = 135$
k=3						1	1		= 1	$\sigma^3 = 40$
k=4								1	= 1	$\sigma^4 = 50$
Cost.	20	10	135	75	105	40	20	50		
Variable	f_1	$f_2 = 1$	$f_3 = 1/3$	$f_4 = 1/3$	$f_5 = 1/3$	f_6	$f_7 = 1$	$f_8 = 1$		

MCF Optimality Conditions

- For each $p \in P^k$, for each k , the reduced cost c'_p :
 - $c'_p = (d_k c_p + \sum_{ij \in A} d_k \pi_{ij} \delta_{ij}^p) - \sigma^k = \sum_{ij} (d_k c_{ij}^k + d_k \pi_{ij}) \delta_{ij}^p - \sigma^k$
 $\sigma^k = \sum_{ij} (c_{ij}^k + \pi_{ij}) \delta_{ij}^p - \sigma^k / d_k \geq 0$
 - where π, σ are the optimal duals for the current restricted master problem
 - $c'_p = 0$, for each utilized path p implies
 $\sum_{ij} (d_k c_{ij}^k + d_k \pi_{ij}) \delta_{ij}^p = \sigma^k$
 or equivalently,
 $\sum_{ij} (c_{ij}^k + \pi_{ij}) \delta_{ij}^p = \sigma^k / d_k$
 - So if, $\min_{p \in P(k)} c'_p = \sum_{ij} (c_{ij}^k + \pi_{ij}) \delta_{ij}^{p^*} - \sigma^k / d_k \geq 0$, the current solution to the restricted master problem is optimal for the original problem
 - If $\min_{p \in P(k)} c'_p = \sum_{ij} (c_{ij}^k + \pi_{ij}) \delta_{ij}^{p^*} - \sigma^k / d_k < 0$, add p^* to restricted master problem

Data Set

- Data Set

Nodes		807
Links		1,363
	capacitated	292
	uncapacitated	1,071
O/D		17,539
	# Origin	136

- Constraint Matrix Size

			Improvement
	row	column	new_row
Node_Arc	14,155,336	23,905,657	-
Path	18,902	-	17,832
Sub-network	1,499	-	428

Computational Results

- Number of Nodes: 807
- Number of Links: 1,363
- Number of Commodities: 17,539
- Computational Result (IBM RS6000, Model 370)
 - Path Model: 44 minutes
 - Sub-network Model: < 1 minute

Conclusions I

- Choose your formulation carefully
 - Trade-off memory requirements and solution time
 - Sub-network formulation can be effective when low level of congestion in the network
- Problem size often mandates use of combined column and row generation

Conclusions II

- Solution time is affected dramatically by
 - The complexity of the pricing problem
 - Exploitation of problem structure, pre-processing, LP solver selection, etc.