

Generalized Assignment Problems

n jobs and m machines

s_{ij} : size of job/item j on machine/bin i

Assignment: A function $f: \text{jobs} \rightarrow \text{machines}$. $f(j)$ is the machine to which job j is assigned

Minimum makespan: find f to minimize maximum load:

$$\max_i \sum_{j: f(j) = i} s_{ij}$$

Note: problem also called *unrelated machine scheduling*

Generalized Assignment Problems

n jobs and m machines

s_{ij} : size of job/item j on machine/bin i

b_i : capacity of machine/bin i

c_{ij} : cost of assigning item i to bin j

Assignment f is *feasible* if for all i , $\sum_{j: f(j) = i} s_{ij} \leq b_i$

Minimum cost assignment: find a feasible f to minimize

$$\text{cost}(f) = \sum_j c_{f(j),j}$$

Generalized Assignment Problems

n jobs and m machines

s_{ij} : size of job/item j on machine/bin i

b_i : capacity of machine/bin i

p_{ij} : profit for assigning item i to bin j

Assignment f is *feasible* if for all i , $\sum_{j: f(j)=i} s_{ij} \leq b_i$

Maximum profit assignment: find a feasible f to maximize

$$\text{profit}(f) = \sum_j p_{f(j),j}$$

Note: can assume dummy machine so that all jobs are assigned

Assignment polytope

x_{ij} : variable to indicate if j is assigned to i

Assignment polytope:

$$\sum_i x_{ij} \leq 1 \quad \text{for each } j$$

$$\sum_j s_{ij} x_{ij} \leq b_i \quad \text{for each } i$$

Easy case: when $s_{ij} \in \{0, 1, \infty\}$, assignment polytope has integral vertices (adjacency matrix of a bipartite graph which is TUM). All variants can be solved exactly

Solvability of GAP

- For general values of s_{ij} , it is NP-Complete to check if there exists a feasible assignment satisfying capacity constraints
- implies cost minimization is inapproximable
- Also NP-hard to minimize makespan to better than a factor of $3/2$
- APX-hard to maximize profit

Above results hold for $s_{ij} \in \{0, 1, 2, \infty\}$

Approximation for GAP

We will show

- 2-approx for minimizing makespan
- a bi-criteria approx for cost minimization, that is we find an assignment f such that $\text{cost}(f) \leq \text{cost}(f^*)$ and for all i ,
$$\sum_{j: f(j) = i} s_{ij} \leq 2 b_i$$
- 1/2-approx for maximizing profit

Minimizing makespan/unrelated machine scheduling

Natural relaxation:

$$\begin{aligned} \min z \\ \sum_j s_{ij} x_{ij} &\leq z \quad \text{for each machine } i \\ \sum_i x_{ij} &= 1 \quad \text{for each job } j \\ x_{ij} &\geq 0 \end{aligned}$$

Has unbounded integrality gap!

Minimizing makespan

Integrality gap example

1 job with $s_{i1} = 1$ for each i

Feasible solution to LP:

$$x_{i1} = 1/m$$

$$z = 1/m$$

$$\text{OPT} = 1$$

so gap = m

Improving gap

Guess value of **OPT**, say **T**

set $x_{ij} = 0$ if $s_{ij} > T$

LP relaxation with parameter **T**

LP(T):

$\sum_j s_{ij} x_{ij} \leq T$ for each machine **i**

$\sum_i x_{ij} = 1$ for each job **j**

$x_{ij} = 0$ if $s_{ij} > T$

$x_{ij} \geq 0$ if $s_{ij} \leq T$

Improving gap

$$T^* = \inf_T (\text{LP}(T) \text{ is feasible})$$

$$\text{Claim: } \text{OPT} \geq T^*$$

T^* can be found by binary search

$$\text{Theorem: } \text{OPT} \leq 2T^*$$

Rounding of LP(T)

$$\sum_j s_{ij} x_{ij} \leq T \quad \text{for each machine } i$$

$$\sum_i x_{ij} = 1 \quad \text{for each job } j$$

$$x_{ij} = 0 \quad \text{if } s_{ij} > T$$

$$x_{ij} \geq 0 \quad \text{if } s_{ij} \leq T$$

$$A_i = \{ j \mid s_{ij} \leq T \}: \text{ allowed items for } i$$

Let x^* be a *basic feasible solution* (extreme point/vertex)
to LP(T)

Properties of basic solutions

a basic solution is defined by k constraints where k is total number of variable (here $k = nm$)

of non-trivial constraints in system is $n+m$ (one for each job and one for each machine)

therefore in x^* , the number of strictly positive variables ($x_{ij}^* > 0$) is at most $n+m$

$F = \{ j \mid x_{ij}^* \neq 1 \text{ for any } i \}$

$j \in F$ implies j is assigned fractionally to more than one machine

Claim: $|F| \leq m$

Properties of basic solutions

of non-trivial constraints in system is $n+m$ (one for each job and one for each machine)

therefore in x^* , the number of strictly positive variables ($x_{ij}^* > 0$) is at most $n+m$

Claim: $|F| \leq m$

Counting: if $j \notin F$ then exactly one i s.t. $x_{ij}^* > 0$
if $j \in F$ then at least two i s.t. $x_{ij}^* > 0$

therefore $(n - |F|) + 2|F| \leq n + m$

implies $|F| \leq m$

Properties of basic solutions

Can say something stronger

Create a bipartite graph $G = (M, F, E)$ where F is fractionally assigned jobs, M is machines and $(i,j) \in E$

iff $x_{ij}^* > 0$

Let G_1, G_2, \dots, G_k be *connected components* of G

with $G_l = (M_l, F_l, E_l)$

Claim: $|F_l| \leq |M_l|$

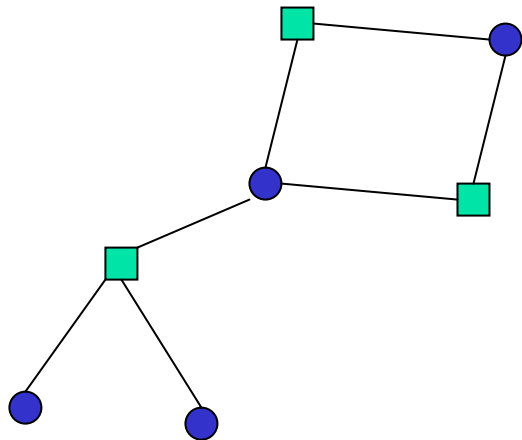
Exercise using same argument as previous claim

Rounding x^*

hence graph G is a *pseudo-forest*: collection of *pseudo-trees*

pseudo-tree is a tree + one edge

Note: each job in G has $\text{deg} \geq 2$



■ job

● machine

Rounding x^*

Process G as follows:

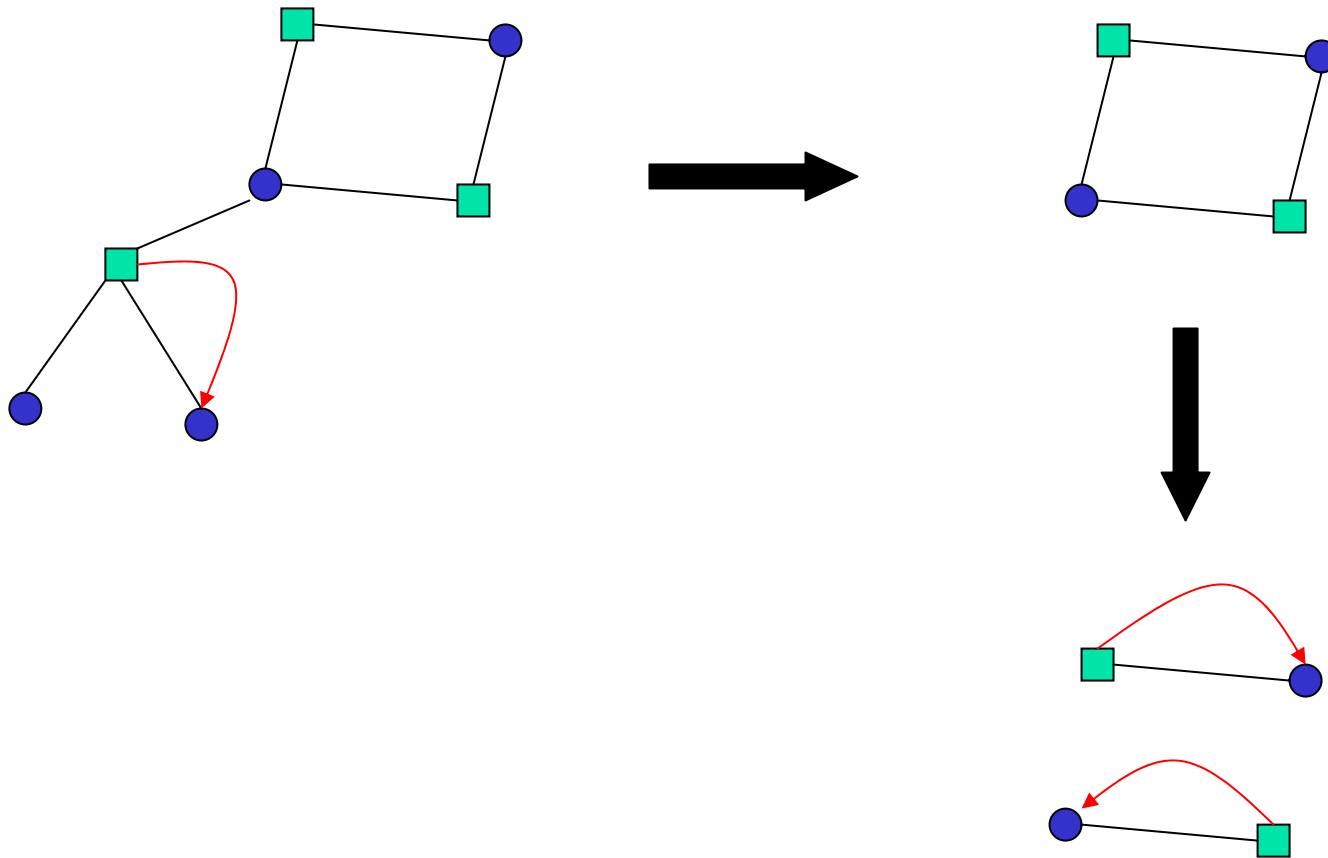
if there is a machine i s.t $\deg(i) = 1$ in G then assign the unique job j with $x_{ij}^* > 0$ to i and remove i, j from G

if no machine i with $\deg(i) = 1$ then G is a collection of cycles

find an arbitrary **maximum matching** in each cycle and assign jobs according to matching

since cycle is even all jobs in cycle assigned

Example



Analysis

Easy to see that each job in F assigned

Further a machine i is assigned at most one job j from F s.t. $x_{ij}^* > 0$ (which implies $s_{ij} \leq T$)

Final load on machine i ?

$$\leq T + \sum_{j: x_{ij}^* = 1} s_{ij}$$

$$\leq T + \sum_j s_{ij} x_{ij}^*$$

$$\leq T + T$$

$$\leq 2T$$

Analysis

Analysis shows that if x^* is a feasible soln to

$$\sum_i x_{ij} = 1$$
$$\sum_j s_{ij} x_{ij} \leq b_i$$

$$x_{ij} = 0 \text{ if } s_{ij} > b_i$$
$$x_{ij} \geq 0 \text{ if } s_{ij} \leq b_i$$

Then there is a $0,1$ soln x' s.t

$$\sum_i x_{ij} = 1$$

$$\sum_j s_{ij} x'_{ij} \leq 2b_i$$

and $x'_{ij} > 0$ implies $x^*_{ij} > 0$

Cost minimization

$$\min \sum_{ij} c_{ij} x_{ij}$$

s.t

$$\sum_i x_{ij} = 1$$

$$\sum_j s_{ij} x_{ij} \leq b_i$$

$$x_{ij} = 0 \text{ if } s_{ij} > b_i$$

$$x_{ij} \geq 0 \text{ if } s_{ij} \leq b_i$$

Given a feasible soln x^* to above LP we can round x^* to a 0,1 soln x' s.t

Cost minimization

Given a feasible soln x^* to above LP we can round x^* to a $0,1$ soln x' s.t

$$\sum_j s_{ij} x'_{ij} \leq 2 b_i$$

and

$$\sum_j c_{ij} x'_{ij} \leq \sum_j c_{ij} x^*_{ij}$$

and

$$x'_{ij} > 0 \text{ implies } x^*_{ij} > 0$$

Rounding

Rounding is more involved than for makespan minimization. Need to preserve cost. Easy to check that previous rounding might violate cost by large amount.

We set $x'_{ij} = x^*_{ij}$ if $x^*_{ij} \in \{0, 1\}$

As before F is set of fractionally assigned items, that is $F = \{j \mid x^*_{ij} \neq 1 \text{ for any } i\}$

We can focus on F

Rounding

Reduce problem to the case where $s_{ij} \in \{0, 1, \infty\}$, the simple assignment problem and use the integrality properties of the assignment polytope

For each bin i , let $n_i = \lceil \sum_j x_{ij}^* \rceil$, the ceiling of number of items assigned to i

We create a new instance in which we replace machine i by n_i new bins

Rounding

New instance. Same items as before, but $\sum_i n_i$ bins. An item j that could have been assigned to an original bin i can be assigned to any copy i' of bin i . The cost of assigning to any copy of i is the same as assigning to i

We no longer have sizes - only have assignment constraints

Construct a feasible solution y to the new assignment problem carefully as follows

Setting up y

Consider bin i with $k=n_i$ copies i_1, i_2, \dots, i_k

Let $A_i = \{j \mid s_{ij} \leq b_i\}$

Note that $x_{ij}^* > 0$ implies $j \in A_i$

Assume wlog that items are numbered s.t
 $s_{i1} \geq s_{i2} \geq \dots s_{ih}$ where $h = |A_i|$

By defn $\lceil \sum_{j=1}^h x_{ij}^* \rceil = n_i$

Setting up y

Consider bin i with $k=n_i$ copies i_1, i_2, \dots, i_k
Let $A_i = \{j \mid s_{ij} \leq b_i\}$

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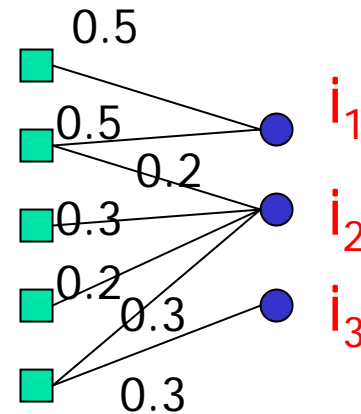
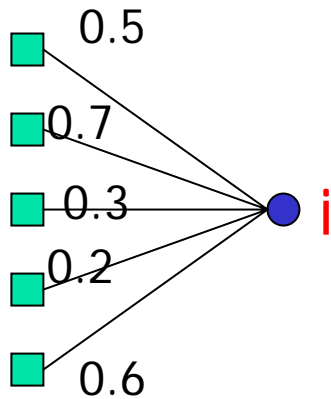
We explain how to set y using an example.

$h = 5$ with $x^*_{ij} = 0.5, 0.7, 0.3, 0.2, 0.6, j = 1$ to 5

Setting up y

We explain how to set y using an example.

$h = 5$ with $x^*_{ij} = 0.5, 0.7, 0.3, 0.2, 0.6, j = 1$ to 5



Setting up y

We ensure that y assigns to each copy of i a fractional value of at most 1 item

More formally the assignment is computed as follows
(easier to understand informally from example)

For $j = 1$ to h

let $z_j = \sum_{i < j} x^*_{ij}$ and let $a = \lfloor z_j \rfloor$

Set $y_{i_{a+1}j} = \min(x^*_{ij}, a+1 - z_j)$

if $x^*_{ij} > y_{i_{a+1}j}$ then set $y_{i_{a+2}j} = x^*_{ij} - y_{i_{a+1}j}$

Properties of y

By construction each new bin is fractionally assigned at most 1 item in y

For item j , x_{ij}^* is distributed to copies of bin i (at most two copies)

Important property: each job fractionally assigned to i_l in y has size (in i) no larger than any job fractionally assigned to i_{l-1}

The cost of y is same as cost of x^*

Therefore y defines a feasible solution to the simple assignment problem

Rounding y to y'

y is a feasible solution to an assignment problem

By integrality of the assignment polytope there *exists* a
integral solution y' s.t

$$\text{cost}(y') \leq \text{cost}(y)$$

$$y'_{i_a j} = 1 \text{ implies } y_{i_a j} > 0$$

at most one item is assigned to each bin

Such a y' can also be obtained in polynomial time from y
using simple operations.

Using y' to obtain x'

If an item j is assigned to a copy of bin i in y' we set

$$x'_{ij} = 1$$

We claim that $\text{cost}(x') \leq \text{cost}(x^*)$

and

$$\sum_j s_{ij} x'_{ij} \leq 2 b_i \text{ for each } i$$

The first part is easy since $\text{cost}(x') = \text{cost}(y') \leq \text{cost}(y) = \text{cost}(x^*)$

Load property of x'

Consider any bin i with copies i_1, i_2, \dots, i_k

Let j_l be the item assigned to copy i_l by y'

(assume for simplicity that i_k is also assigned an item
though it is not always the case)

for $l = 2$ to k we claim that $s_{ij_l} \leq \sum_j s_{ij} y_{i_{l-1},j}$

This follows from the ordering of jobs according to sizes in
construction y from x^*

Therefore $\sum_{l=2}^k s_{ij_l} \leq \sum_{l=2}^k \sum_j s_{ij} y_{i_{l-1},j}$

Load property of x'

Therefore

$$\sum_{l=2}^k s_{ij_l} \leq \sum_{l=2}^k \sum_j s_{ij} y_{i_{l-1},j} \leq \sum_j s_{ij} x_{ij}^* \leq b_i$$

The second inequality follows from the fact x_{ij}^* is spread amongst copies of i by y

Thus the load of items j_2, j_3, \dots, j_k is at most b_i . The load of item j_1 can be at most b_i . Therefore the total load on i is at most $2b_i$

Load property of x'

We charge j_k to fractional load on i_{k-1} , charge j_{k-1} to fractional load on i_{k-1} and so on. The last item j_1 cannot be charged.

Profit maximization

$$\min \sum_{ij} p_{ij} x_{ij}$$

s.t

$$\sum_i x_{ij} \leq 1$$

$$\sum_j s_{ij} x_{ij} \leq b_i$$

$$x_{ij} = 0 \text{ if } s_{ij} > b_i$$

$$x_{ij} \geq 0 \text{ if } s_{ij} \leq b_i$$

Given a feasible soln x^* to above LP we can round x^* to a feasible $0,1$ soln x' s.t $\text{profit}(x') \geq \frac{1}{2} \text{profit}(x^*)$

Note that no bin capacity constraints are violated in x'

Reduction to cost minimization

First by adding an extra dummy bin $m+1$ with infinite capacity and $p_{(m+1),j} = 0$ for all j we can assume that all items should be assigned. That is

$$\min \sum_{ij} p_{ij} x_{ij}$$

s.t

$$\sum_i x_{ij} = 1$$
$$\sum_j s_{ij} x_{ij} \leq b_i$$

$$x_{ij} = 0 \text{ if } s_{ij} > b_i$$

$$x_{ij} \geq 0 \text{ if } s_{ij} \leq b_i$$

Reduction to cost minimization

Since all items are assigned we can reduce profit maximization to cost minimization by setting $c_{ij} = -p_{ij}$

Exercise (hw): show how the algorithm for cost minimization can be used to obtain a *feasible* solution to profit maximization with $\frac{1}{2}$ the optimum profit

Hint: each bin gets only one extra item that violates its capacity