



CS240 Algorithm Design and Analysis

Lecture 27

Approximation Algorithms

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Generalized Load Balancing





Generalized Load Balancing



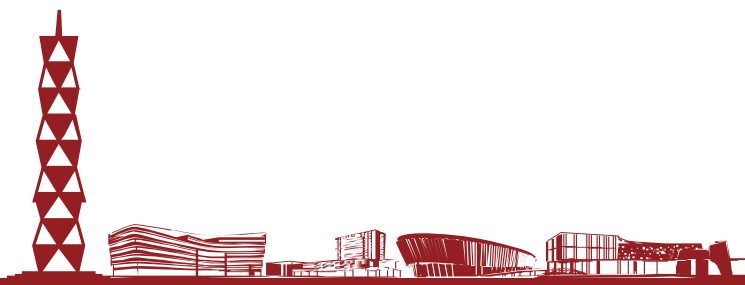
Input. Set of m machines M ; set of n jobs J .

- Job j must run contiguously on an **authorized machine** in $M_j \subseteq M$.
- Job j has processing time t_j .
- Each machine can process at most one job at a time.

Def. Let $J(i)$ be the subset of jobs assigned to machine i . The load of machine i is $L_i = \sum_{j \in J(i)} t_j$.

Def. The makespan is the maximum load on any machine $= \max_i L_i$.

Generalized load balancing. Assign each job to an authorized machine to minimize makespan.





Generalized Load Balancing: Integer Linear Program and Relaxation



ILP formulation. x_{ij} = time machine i spends processing job j .

$$\begin{aligned} (IP) \quad & \min \quad L \\ \text{s. t.} \quad & \sum_i x_{ij} = t_j \quad \text{for all } j \in J \\ & \sum_j x_{ij} \leq L \quad \text{for all } i \in M \\ & x_{ij} \in \{0, t_j\} \quad \text{for all } j \in J \text{ and } i \in M_j \\ & x_{ij} = 0 \quad \text{for all } j \in J \text{ and } i \notin M_j \end{aligned}$$

LP relaxation.

$$\begin{aligned} (LP) \quad & \min \quad L \\ \text{s. t.} \quad & \sum_i x_{ij} = t_j \quad \text{for all } j \in J \\ & \sum_j x_{ij} \leq L \quad \text{for all } i \in M \\ & x_{ij} \geq 0 \quad \text{for all } j \in J \text{ and } i \in M_j \\ & x_{ij} = 0 \quad \text{for all } j \in J \text{ and } i \notin M_j \end{aligned}$$





Generalized Load Balancing: Lower Bounds

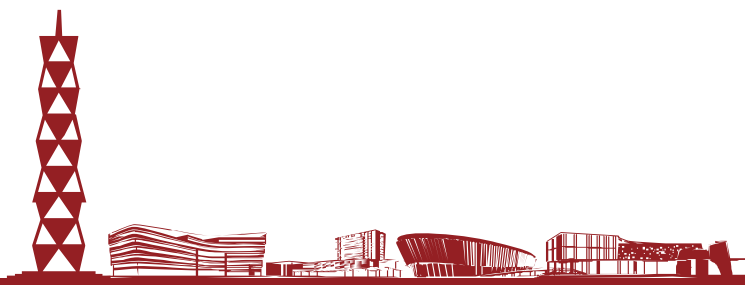


Lemma 1. The optimal makespan $L^* \geq \max_j t_j$.

Pf. Some machine must process the most time-consuming job. ■

Lemma 2. Let L be the optimal value to the LP. Then, the optimal makespan $L^* \geq L$.

Pf. LP has fewer constraints than IP formulation. ■





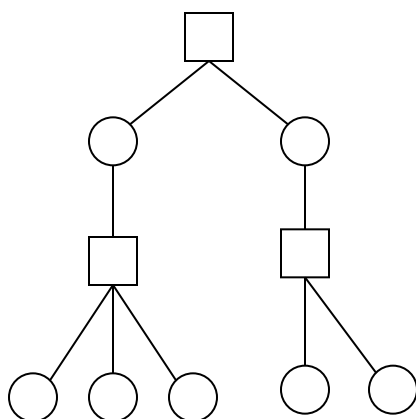
Generalized Load Balancing: structure of LP solution



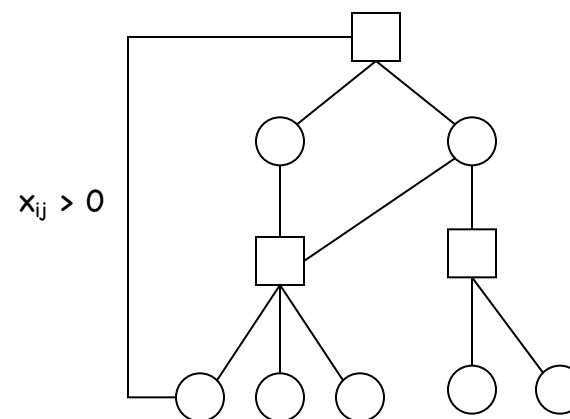
Lemma 3. Let x be solution to LP. Let $G(x)$ be the graph with an edge from machine i to job j if $x_{ij} > 0$. Then $G(x)$ is **acyclic**.

Pf. (deferred)

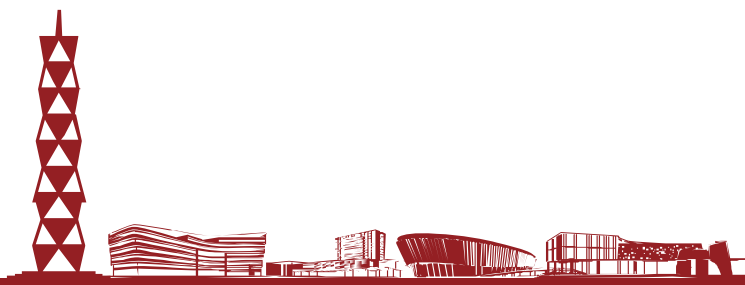
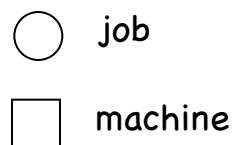
can transform x into another LP solution where $G(x)$ is acyclic if LP solver doesn't return such an x



$G(x)$ acyclic



$G(x)$ cyclic





Generalized Load Balancing: Rounding

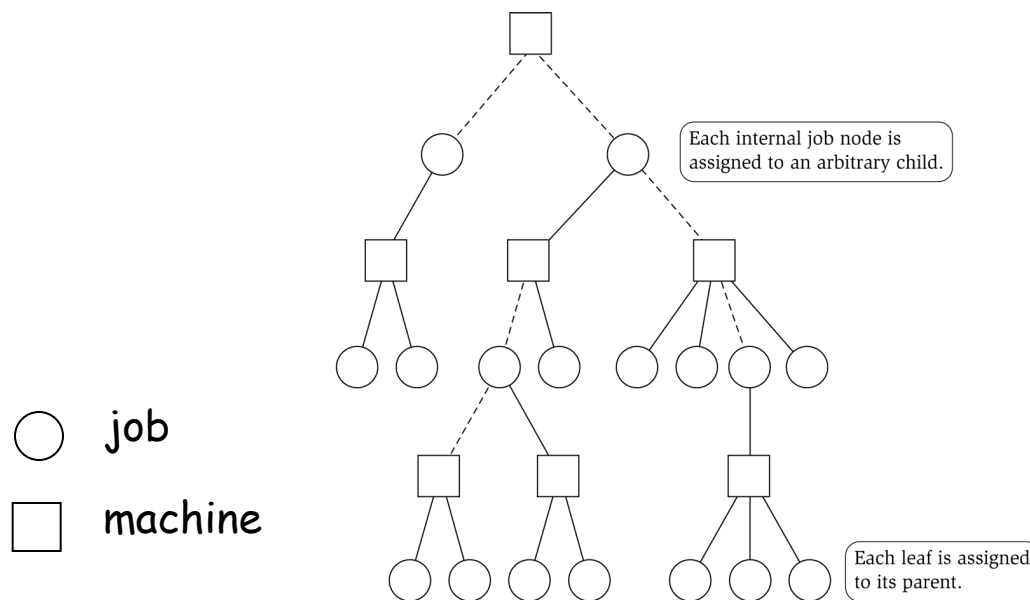


Rounded solution. Find LP solution x where $G(x)$ is a forest. Root forest $G(x)$ at some arbitrary machine node r .

- If job j is a leaf node, assign j to its parent machine i .
- If job j is not a leaf node, assign j to one of its children.

Lemma 4. Rounded solution only assigns jobs to authorized machines.

Pf. If job j is assigned to machine i , then $x_{ij} > 0$. LP solution can only assign positive value to authorized machines. ■





Generalized Load Balancing: Analysis

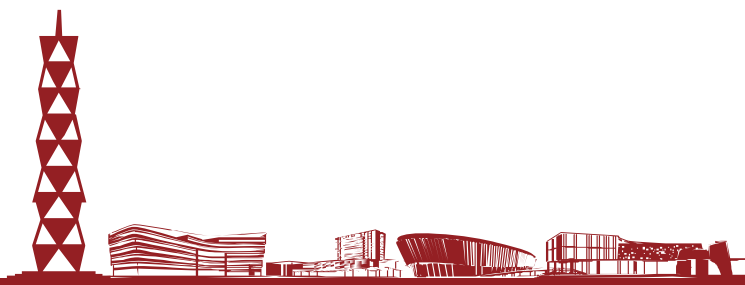
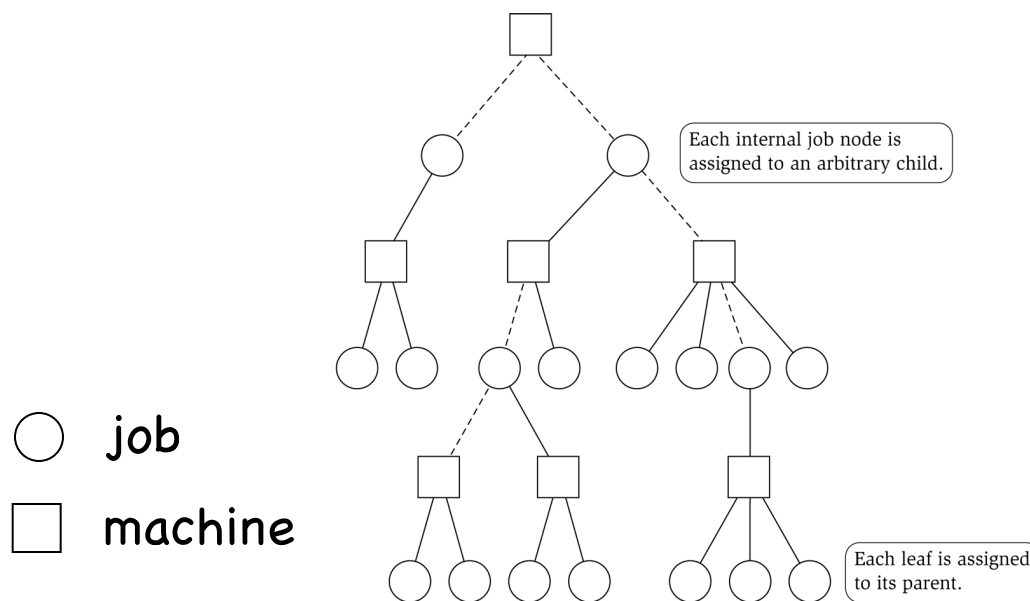


Lemma 5. If job j is a leaf node and machine $i = \text{parent}(j)$, then $x_{ij} = t_j$.

Pf. Since j is a leaf, $x_{ij} = 0$ for all $i \neq \text{parent}(j)$. LP constraint guarantees $\sum_i x_{ij} = t_j$. ■

Lemma 6. At most one non-leaf job is assigned to a machine.

Pf. The only possible non-leaf job assigned to machine i is $\text{parent}(i)$. ■





Generalized Load Balancing: Analysis



Theorem. Rounded solution is a 2-approximation.

Pf.

- Let $J(i)$ be the jobs assigned to machine i .
- By Lemma 6, the load L_i on machine i has two components:

- leaf nodes

- parent(i)

$$\begin{array}{c} \text{Lemma 5} \\ \downarrow \\ \sum_{\substack{j \in J(i) \\ j \text{ is a leaf}}} t_j = \sum_{\substack{j \in J(i) \\ j \text{ is a leaf}}} x_{ij} \leq \sum_{j \in J} x_{ij} \leq L \leq L^* \\ \text{LP} \quad \text{Lemma 2 (LP is a relaxation)} \\ \downarrow \quad \downarrow \\ \text{optimal value of LP} \uparrow \\ \text{Lemma 1} \\ \downarrow \\ t_{\text{parent}(i)} \leq L^* \end{array}$$

- Thus, the overall load $L_i \leq 2L^*$. ■



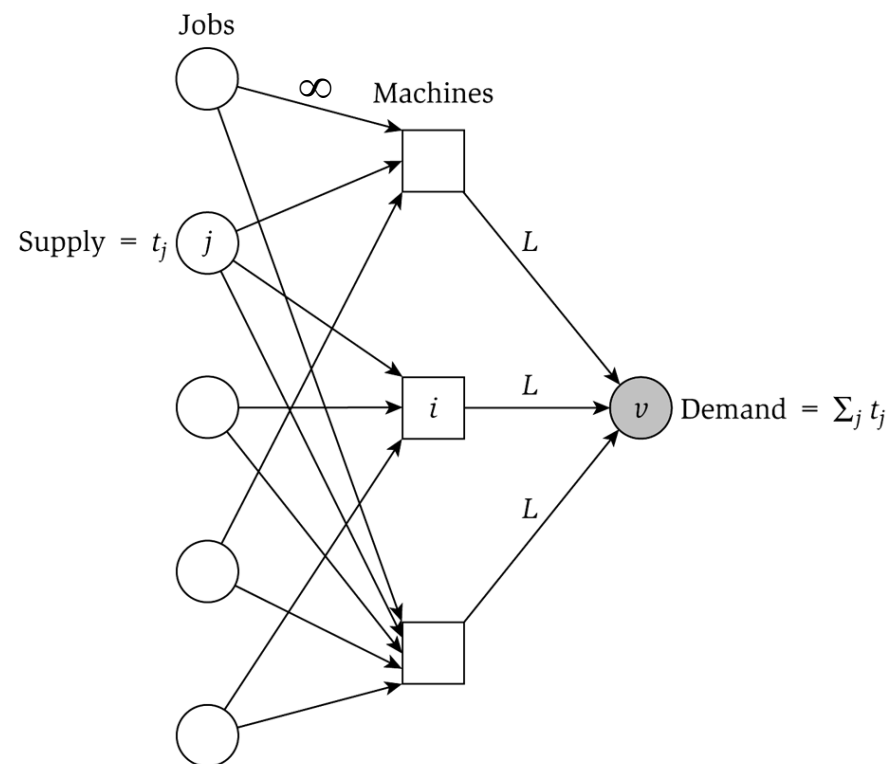


Generalized Load Balancing: Flow Formulation



Flow formulation of LP.

$$\begin{aligned}\sum_i x_{ij} &= t_j && \text{for all } j \in J \\ \sum_j x_{ij} &\leq L && \text{for all } i \in M \\ x_{ij} &\geq 0 && \text{for all } j \in J \text{ and } i \in M_j \\ x_{ij} &= 0 && \text{for all } j \in J \text{ and } i \notin M_j\end{aligned}$$



Observation. Solution to feasible flow problem with value L are in one-to-one correspondence with LP solutions of value L .





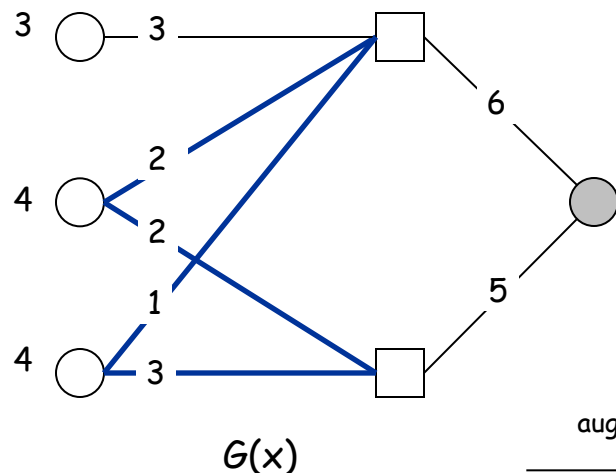
Generalized Load Balancing: Structure of Solution



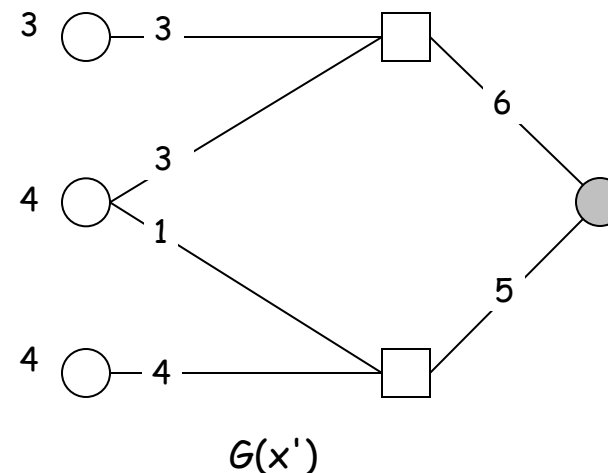
Lemma 3. Let (x, L) be solution to LP. Let $G(x)$ be the graph with an edge from machine i to job j if $x_{ij} > 0$. We can find another solution (x', L) such that $G(x')$ is acyclic.

Pf. Let C be a cycle in $G(x)$.

- Augment flow along the cycle C . ← flow conservation maintained
- At least one edge from C is removed (and none are added).
- Repeat until $G(x')$ is acyclic.

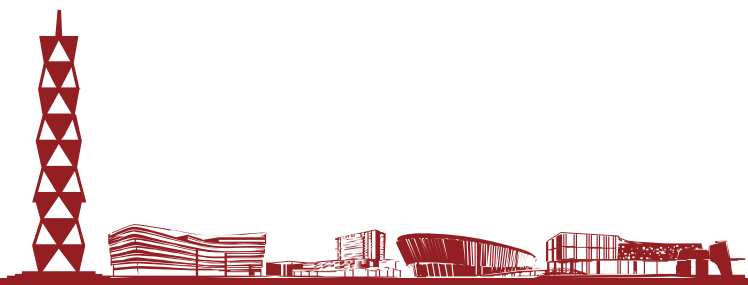


augment along C





K-Center Problem

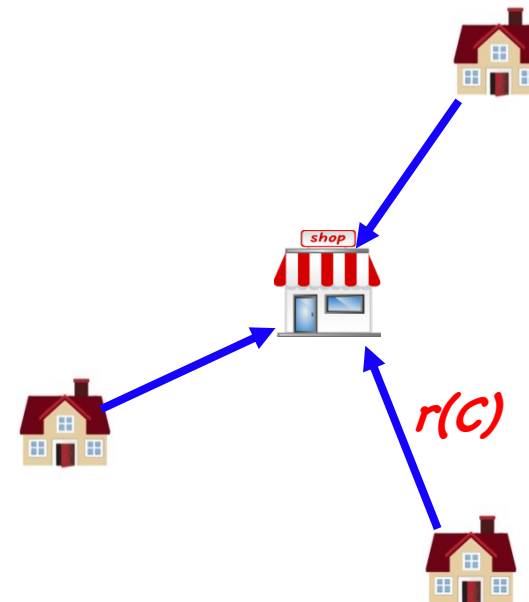
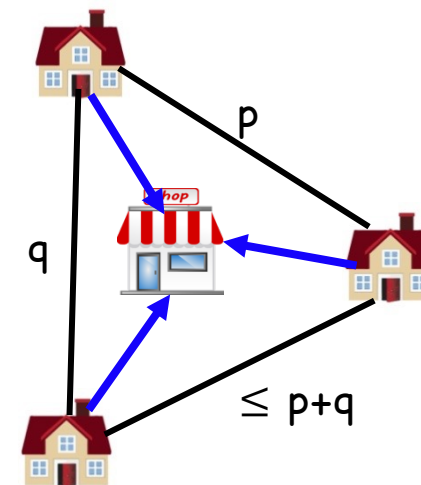




K-Center Problem



- Given a city with n sites, we want to build k centers to serve them.
 - Let S be set of sites, C be set of centers.
- Each site uses the center closest to it.
 - Distance of site s from the nearest center is $d(s, C) = \min_{c \in C} d(s, c)$.
- Goal is to make sure no site is too far from its center.
 - We want to minimize the max distance that any site is from its closest center.
 - Minimize $r(C) = \max_{s \in S} \min_{c \in C} d(s, c)$.
 - C is called a cover of S , and r is called C 's radius.
 - Where should we put centers to minimize the radius?
- Assume distances satisfy triangle inequality.

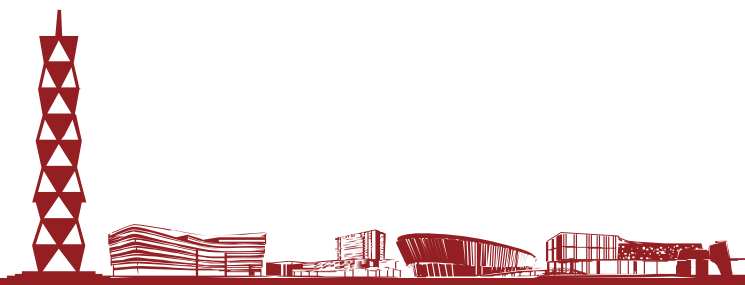
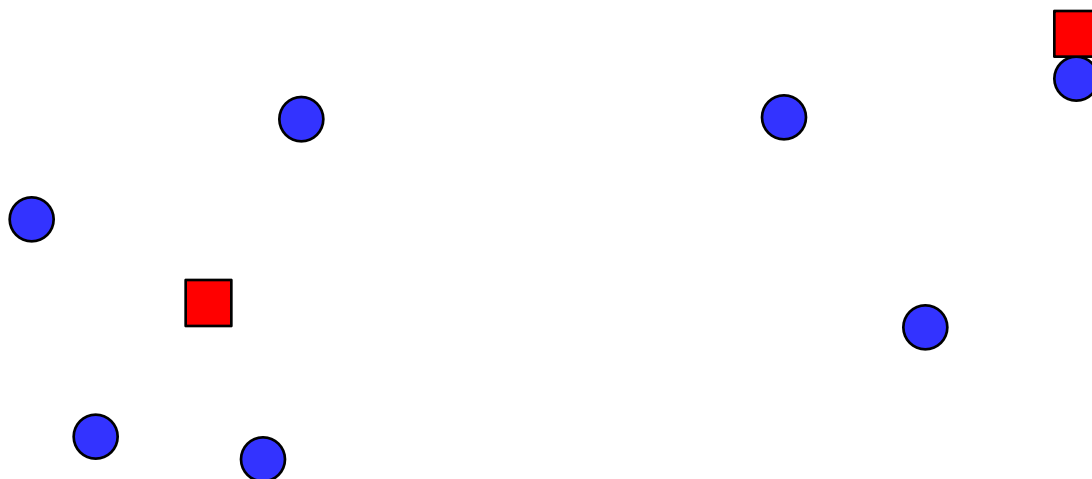




Gonzalez's Algorithm



- k-Center is NP-complete.
- We'll give a simple 2-approximation for it.
- **Idea** Say there's one site that's farthest away from all centers. Then it makes the radius large. We'll put a center at that site, to reduce the radius.
 - Note we allow putting center at same location as site.

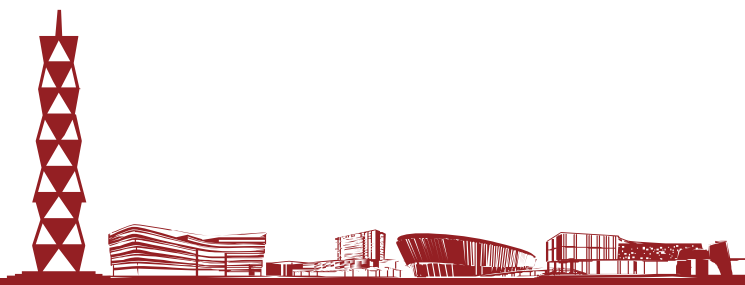




Gonzalez's Algorithm



- C is set of centers, initially empty.
- repeat k times
 - choose site s with maximum $d(s, C)$
 - add s to C
- return C
- **Note** The centers are located at the sites.

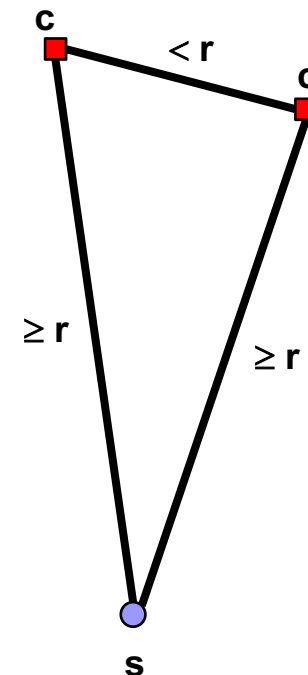




Proof of Correctness



- Let C be the algorithm's output, and r be C 's radius.
 - $r = \max_{s \in S} \min_{c \in C} d(s, c)$
- **Lemma 1** For any $c, c' \in C, d(c, c') \geq r$.
- **Proof** Since r is the radius, there exists a point $s \in S$ at distance $\geq r$ from all the centers.
 - If there's no such s , then C 's radius $< r$.
 - So s is distance $\geq r$ from c and c' .
 - Suppose WLOG c' is added to C after c .
 - If $d(c, c') < r$, then algorithm would add s to C instead of c' , since s is farther.

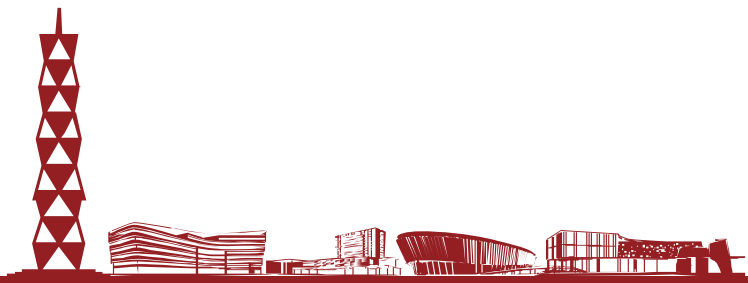




Proof of Correctness



- **Cor** There exist $k+1$ points mutually at distance $\geq r$ from each other.
 - By the lemma, the k centers are mutually $\geq r$ distance apart.
 - Also, there's an $s \in S$ at distance $\geq r$ from all the centers.
 - Otherwise, C 's covering radius is $< r$.
 - So, the k centers plus s are the $k+1$ points.
- Call these $k+1$ points D .

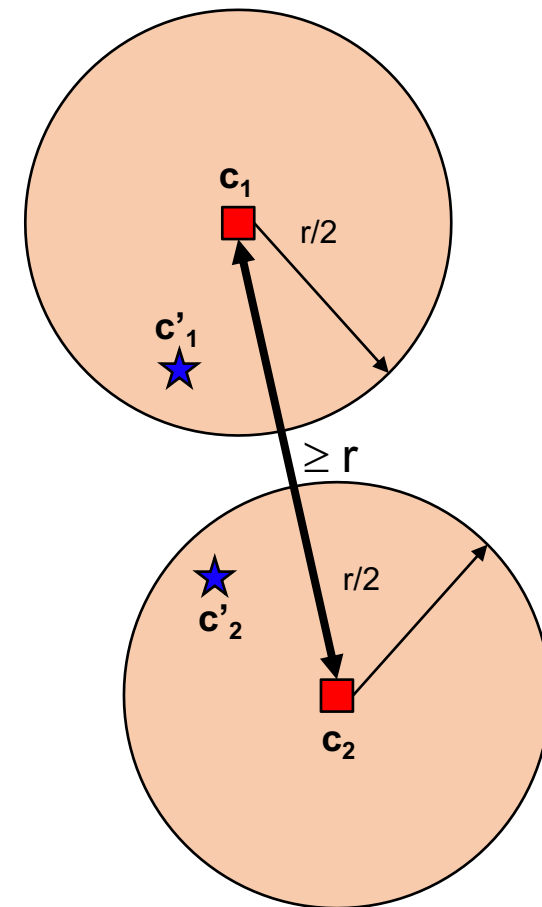




Proof of Correctness



- Let C^* be an optimal cover with radius r^* .
- **Lemma 2** Suppose $r > 2r^*$. Then for every $c \in D$, there exists a corresponding $c' \in C^*$. Furthermore, all these c' are unique.
- **Proof** Draw a circle of radius $r/2$ around each $c \in D$.
 - There must be a $c' \in C^*$ inside the circle, because
 - c is at most distance r^* away from its nearest center, since r^* is C^* 's radius.
 - $r/2 > r^*$.
 - Given $c_1, c_2 \in D$, let $c'_1, c'_2 \in C^*$ be inside c_1 and c_2 's circle, resp.
 - c_1 and c_2 's circles don't touch, because $d(c_1, c_2) \geq r$.
 - So $c'_1 \neq c'_2$





Proof of Correctness



- **Thm** Let C be the output of Gonzalez's algorithm and let C^* be an optimal k -center. Then $r(C) \leq 2r(C^*)$.
- **Proof** By Lemma 2, if $r(C) > 2r(C^*)$, then for every $c \in D$, there is a unique $c' \in C^*$.
 - But there are $k+1$ points in D , by the corollary.
 - So, there are $k+1$ points in C^* . This is a contradiction because C^* is a k -center.

