The Design and Analysis of Algorithms

Lecture 7 Greedy Algorithms II

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Content

Optimal Cashing

Minimum Spanning Tree

Clustering





Optimal Offline Caching

Caching. Cache with capacity to store *k* items.

Sequence of m item requests d_1, d_2, \dots, d_m .

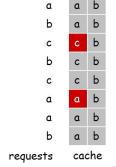
Cache hit: item already in cache when requested.

Cache miss: item not already in cache when requested: must bring requested item into cache, and evict some one, if full.

Goal. Schedule that minimizes number of evictions.

Ex. k = 2, initial cache = ab, requests: a, b, c, b, c, a, a, b.

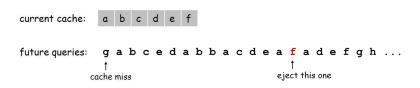
Optimal eviction schedule: 2 evictions.





Optimal Offline Caching: Greedy Algorithms

- LIFO / FIFO. Evict element brought in most/least recently.
- LRU. Evict element whose most recent access was earliest.
- LFU. Evict element that was least frequently requested.
- Farthest-in-future. Evict item in the cache that is not requested until farthest in the future.



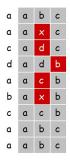
Claim. FF is optimal eviction schedule!

Algorithm is intuitive; proof is subtle.

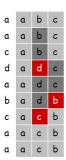


Reduced Eviction Schedules

- Def. A *reduced* schedule is a schedule that only inserts an item into the cache in a step in which that item is requested.
 - Intuition. Can transform an unreduced schedule into a reduced one with no more cache misses.



an unreduced schedule



a reduced schedule

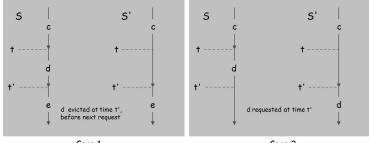


Reduced Eviction Schedules

- Claim. Given any unreduced schedule S, can transform it into a reduced schedule S' with no more evictions.
 - Pf. [by induction on number of unreduced items] Suppose *S* brings *d* into the cache at time *t*, without a request.

Let *c* be the item *S* evicts when it brings *d* into the cache.

- Case 1. d evicted at time t', before next request for d.
- Case 2. d requested at time t' before d is evicted. \square





Farthest-In-Future: Analysis

Theorem 1 (Bélády 1966)

FF is optimal eviction schedule.

- Pf. Follows directly from the following claim.
- Claim. There exists an optimal reduced schedule S that makes the same eviction schedule as S_{FF} through the first j requests.
 - Pf. [by induction on *j*] Let *S* be reduced schedule that satisfies the claim through *j* requests.
 - We produce S' that satisfies the claim after j + 1 requests.





Consider $(j+1)^{st}$ request $d=d_{j+1}$.

Since S and S_{FF} have agreed up until now, they have the same cache contents before request j + 1.

Case 1: [d is already in the cache]. S' = S satisfies the claim.

Case 2: [d is not in the cache and S and S_{FF} evict the same element]. S' = S satisfies the claim.

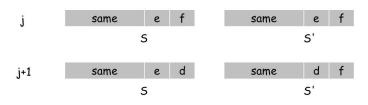


Case 3: [*d* is not in the cache; S_{FF} evicts e; S evicts $f \neq e$].

Construction of S' from S by evicting e instead of f.

S' agrees with S_{FF} on first j+1 requests; we show that having element f in cache is no worse than having element e.

Let S' behave the same as S until S' is forced to take a different action (because either S evicts e; or because either e or f is requested)







Let j' be the first time after j+1 that S' must take a different action from S, and let g be item requested at time j'.

• Case 3a: *g* = e.

Can't happen with FF since there must be a request for f before e.

• Case 3b: g = f.

Element f can't be in cache of S, so let e' be the element that S evicts.

if e' = e, S' accesses f; now S and S' have same cache;

if $e' \neq e$, we make S' evict e' and brings e into the cache; now S and S' have the same cache.





We let S' behave exactly like S for remaining requests.

• Case 3c: $g \neq e, f$. S evicts e.

Make S' evict f.

Now S and S' have the same cache. □

j'	same	g	same	g
	S		S'	

Caching Perspective

Online vs. offline algorithms.

Offline: full sequence of requests is a priori.

Online (reality): requests are not known in advance.

Caching is among most fundamental online problems in CS.

- LIFO. Evict page brought in most recently.
- LRU. Evict page whose most recent access was earliest.

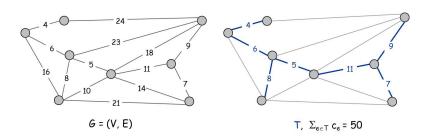
LRU is *k*-competitive. [Section 13]

LIFO is arbitrarily bad.



Minimum Spanning Tree

Minimum spanning tree. Given a connected graph G = (V, E) with real-valued edge weights c_e, an MST is a subset of the edges T ⊆ E such that T is a spanning tree whose sum of edge weights is minimized.



Theorem 2 (Cayley's Theorem)

There are n^{n-2} spanning trees of K_n .





Applications

MST is fundamental problem with diverse applications.

Network design.

Approximation algorithms for NP-hard problems.

Max bottleneck paths.

Cluster analysis.

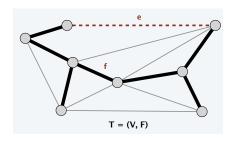




Fundamental Cycle

- Adding any non-tree edge e to a spanning tree T forms unique cycle C.
- Deleting any edge $f \in C$ from $T \cup \{e\}$ results in new spanning tree.

Observation. If $c_e < c_f$, then T is not an MST.

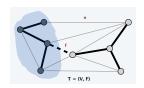






Fundamental Cutset

- Deleting any tree edge f from a spanning tree T divide nodes into two connected components. Let D be cutset.
- Adding any edge e ∈ D to T {f} results in new spanning tree.
 Observation. If c_e < c_f, then T is not an MST.



Greedy Algorithms

- Kruskal's algorithm. Start with $T = \emptyset$. Consider edges in ascending order of cost. Insert edge e in T unless doing so would create a cycle.
- Prim's algorithm. Start with some root node s and greedily grow a tree T from s outward. At each step, add the cheapest edge e to T that has exactly one endpoint in T.

Theorem 3

Kruskal's (Prim's) algorithm can find MST in $O(m \log n)$ time.



Clustering

- Clustering. Given a set U of n objects labeled p_1, \dots, p_n , classify into coherent groups.
- Distance function. Numeric value specifying "closeness" of two objects.
- Fundamental problem. Divide into clusters so that points in different clusters are far apart.

Routing in mobile ad hoc networks.

Identify patterns in gene expression.

Document categorization for web search.

Skycat: cluster 10⁹ sky objects into stars, guasars, galaxies.



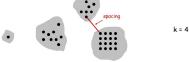


Clustering of Maximum Spacing

- *k-clustering*. Divide objects into *k* non-empty groups.
- Distance function. Assume it satisfies several natural properties.

$$d(p_i, p_j) = 0$$
 iff $p_i = p_j$ (indiscernible)
 $d(p_i, p_j) \ge 0$ (nonnegativity)
 $d(p_i, p_j) = d(p_j, p_i)$ (symmetry)

- Spacing. Min distance between any pair of points in different clusters.
- Clustering of maximum spacing. Given an integer k, find a k-clustering of maximum spacing.





Greedy Clustering Algorithm

Single-link k-clustering algorithm.

Form a graph on the vertex set *U*, corresponding to *n* clusters.

Find the closest pair of objects such that each object is in a different cluster, and add an edge between them.

Repeat n - k times until there are exactly k clusters.

 Key observation. This procedure is precisely Kruskal's algorithm (except we stop when there are k connected components).

Remark. Equivalent to finding an MST and deleting the k-1 most expensive edges.





Greedy Clustering Algorithm: Analysis

Theorem 4

Let C^* denote the clustering C_1^*, \dots, C_k^* formed by deleting the k-1 most expensive edges of a MST. C^* is a k-clustering of max spacing.

Pf. Let C denote some other clustering C_1, \dots, C_k .

The spacing of C^* is the length d^* of the $(k-1)^{st}$ most expensive edge.

Let p_i , p_j be in the same cluster in C^* , say C_r^* , but different clusters in C, say C_s and C_t .

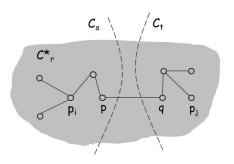




Some edge (p, q) on $p_i - p_j$ path in C_r^* spans two different clusters in C.

All edges on $p_i - p_j$ path have length $\leq d^*$ since Kruskal chose them.

Spacing of C is $\leq d^*$ since p and q are in different clusters. \square







Homework

- Read Chapter 4 of the textbook.
- Exercises 8 & 18 in Chapter 4.

