CS345

## Compact Skeletons

## Compact Skeletons

- Assume tuples components are scattered over website
- We have a tagger that can tag all tuple components on website
- Assume no noise for now
- Reconstruct relation


## Compact Skeletons

Relation
$\uparrow$
Skeleton
1
Data Graph


Website



Address ( $A$ )



| $T$ | $S$ | $D$ | $A$ |
| :--- | :--- | :--- | :--- |
| Programmer | 100 K | R \&D | 1200 Jose Blvd |
| CTO | 150 K | R \& D | 1200 Jose Blvd |
| Admin Asst | 60K | Corporate | $4007^{\text {th }}$ Ave |
| CEO | (null) | Corporate | $4007^{\text {th }}$ Ave |



| $T$ | $S$ | $D$ | $A$ |
| :--- | :--- | :--- | :--- |
| Programmer | 100 K | R \&D | 1200 Jose Blvd |
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## Relation



Skeleton I
Data Graph i
Website

## Skeletons

- Labeled trees
- Transformation from data graphs to relations



## Overlays



## Overlays



| $T$ | $S$ | $D$ | $A$ |
| :---: | ---: | ---: | :---: |
| Programmer | 100K | R \&D | 1200 Jose Blvd |

## Overlays



| $T$ | $S$ | $D$ | $A$ |
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## Overlays



Programmer 100K CTO 150K

## Overlays



## Overlays



| $T$ | $S$ | $D$ | $A$ |
| :--- | ---: | :--- | :---: |
| Programmer | 150 K | $\mathrm{R} \& \mathrm{D}$ | 1200 Jose Blvd |
| CTO | 100 K | R \& D | 1200 Jose Blvd |

## Inconsistent Overlays



Programmer 100K CTO 150K

## Inconsistent Overlays



Programmer 100K CTO 150K

## Compact Skeletons

- A skeleton is compact if all overlays are consistent
- Perfect if each node and edge of data graph is covered by at least one overlay
- Given a data graph G, does G have a Perfect Compact Skeleton (PCS)?
- Not always
- But if it exists it is unique


## PCS Algorithm



Programmer 100K CTO 150K

## PCS Algorithm



Work bottom-up:
Compute node signatures
Place nodes in equivalence classes based on signature Construct skeleton from equivalence classes

PCS Algorithm


## Incomplete information



## Incomplete information



| $T$ | $S$ | $D$ | $A$ |
| :--- | :--- | :--- | :---: |
| Admin Asst | 60 K | Corporate | $4007^{\text {th }}$ Ave |

## Incomplete information



| $T$ | $S$ | $D$ | $A$ |
| :--- | :--- | :--- | :---: |
| Admin Asst | 60 K | Corporate | $4007^{\text {th }}$ Ave |
| CEO | $\perp$ | Corporate | $4007^{\text {th }}$ Ave |

## Partial Compact Skeletons

- For data graphs with incomplete information, we allow partial overlays
- Results in nulls in relation
- If we can use consistent partial overlays to cover every node and edge of the graph, we have a partially perfect compact skeleton (PPCS)


## Tuple subsumption

- Tuple $\boldsymbol{t}$ subsumes tuple $u$ if $t$ and $u$ agree on every component of $u$ that is not null

$$
\begin{array}{llll} 
& T & S & D
\end{array} A
$$

## Noisy Data Graphs

- Real-life websites are noisy
- False positives e.g., MS = degree, state or Microsoft?
- Non-skeleton links e.g., featured products


## Data graph for a retail website



C: Category
I: Item
P: Price
A: Availability

For simplicity: assume all nodes have a label

## Coverage of a skeleton



## Coverage of a skeleton



## Coverage of a skeleton



## Skeletons for Noisy Data Graphs

- Problem:
- Find skeleton K with optimal coverage, called the best-fit skeleton (BFS)
- NP-complete


## Greedy Heuristic for BFS



## Greedy Heuristic for BFS



| Label | Parent | Count |
| :---: | :---: | :---: |
| P | I | 3 |
| A | I | 3 |
|  | C | 1 |
| I | C | 4 |
|  | R | 1 |
| C | R | 1 |
| R O |  |  |
| C ${ }^{\circ}$ |  |  |
| $\begin{aligned} & \mathrm{I} \\ & \mathrm{O} \\ & \mathrm{P} \\ & \mathrm{~A} \end{aligned}$ |  |  |



| Label | Parent | Count |
| :---: | :---: | :---: |
| D | C | 4 |
| $\mathbf{C}$ | $\mathbf{A}$ | 2 |
|  | B | 1 |
| $\mathbf{A}$ | R | 1 |
| $\mathbf{B}$ | $\mathbf{R}$ | 1 |



Greedy skeleton


Greedy skeleton
Coverage $=9$


Greedy skeleton
Coverage $=9$

Optimal skeleton
Coverage $=15$

## Weighted Greedy Heuristic

- Simple Greedy heuristic uses parent counts
- "Memory-less"
- Weighted Greedy heuristic takes into account past selections to improve simple greedy selection
- Computes "benefit" of each decision at every stage



## Weighted Greedy



Greedy skeleton
Coverage = 9


## Weighted Greedy benefit $(\mathrm{A} \rightarrow \mathrm{C})=4$



Greedy skeleton Coverage $=9$


## Weighted Greedy benefit $(\mathrm{A} \rightarrow \mathrm{C})=4$ benefit $(\mathrm{B} \rightarrow \mathrm{C})=10$



Greedy skeleton Coverage $=9$


## Weighted Greedy



Greedy skeleton
Coverage $=9$



Greedy skeleton Coverage $=9$


Weighted greedy skeleton Coverage $=15$

## Summary

Relation
$\uparrow$
Compact Skeleton
I
Data Graph t

Website

