

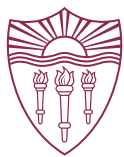


GRAMS: A Graph-based Approach for Inferring Semantic Descriptions of Wikipedia Tables

Binh Vu, Craig Knoblock, Pedro Szekely, Jay Pujara,
Minh Pham

Information Sciences Institute
University of Southern California

Motivating Example



- Wikipedia has 7.5 millions tables covering many domains

List of players won [Walter Payton Award](#)

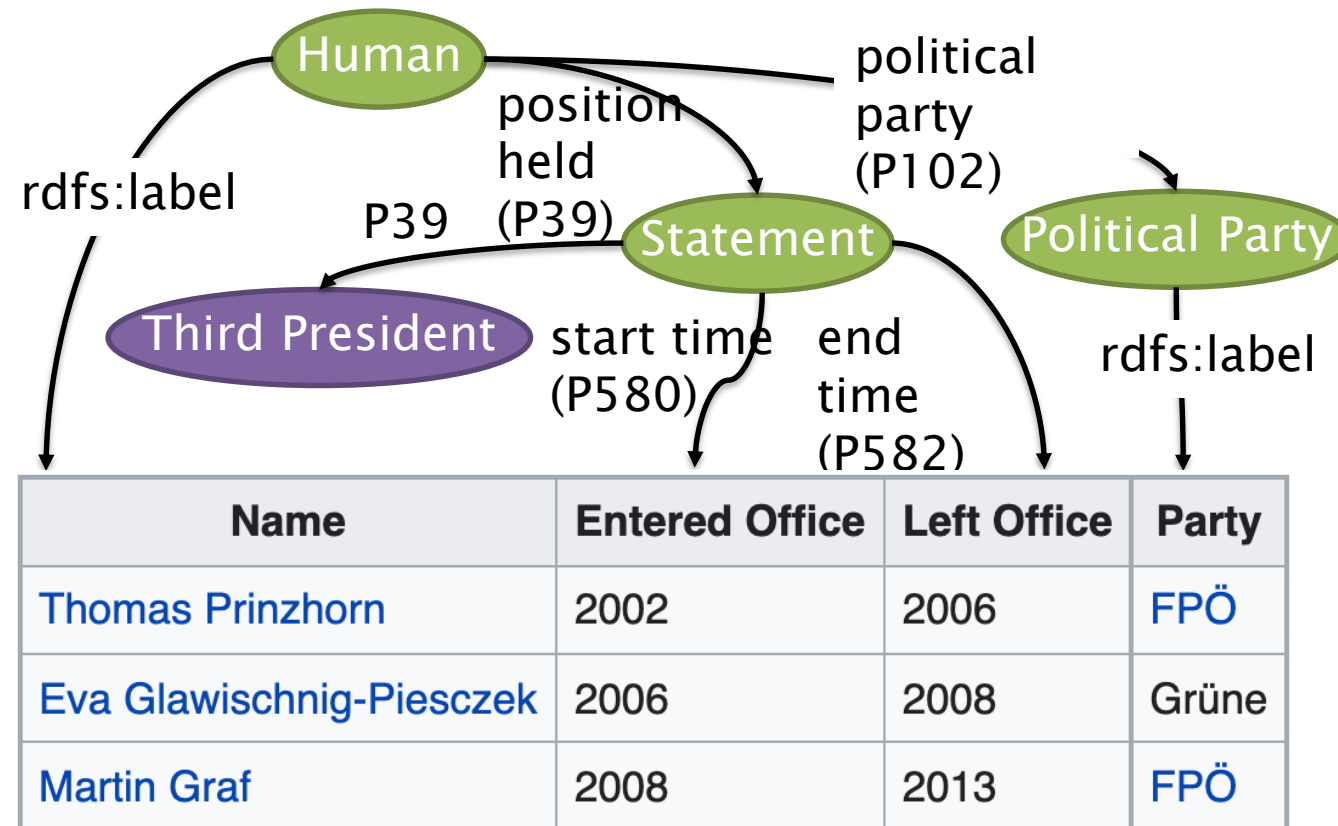
Members of 56th New Brunswick Legislature

Members of 56th New Brunswick Legislature					List of players won Walter Payton Award			
Name		Party		Ri	Year	Player	School	Position
Hédard Albert		Liberal		Caraquet	1987	Kenny Gamble	Colgate	RB
David Alward		Progressive Conservative		Woodstock	1988	Dave Meggett	Towson State	RB
Donald Arseneault		Liberal		Dalhousie-Restigouche East	2018	Devlin Hodges	Samford	QB
John Betts		Progressive Conservative		Moncton Crescent	2019	Trey Lance	North Dakota State	QB
Dereham	Dereham	Inmaze		Regeneron	Ebola Virus			
Aldeby	Aldeby	Olorofim		F2G	invasive mold infections			
Ashwellthorpe	Ashwellthorpe	1881	1939	Great Eastern				
Hotel Diablo		School		Rail	Shellharbour		1959	Website
FLYGOD is an Awesome GOD		westside Gunn		Hip hop				

Source Modeling Problem



- Building semantic descriptions of tables
 - Describing data source using classes and properties in ontologies

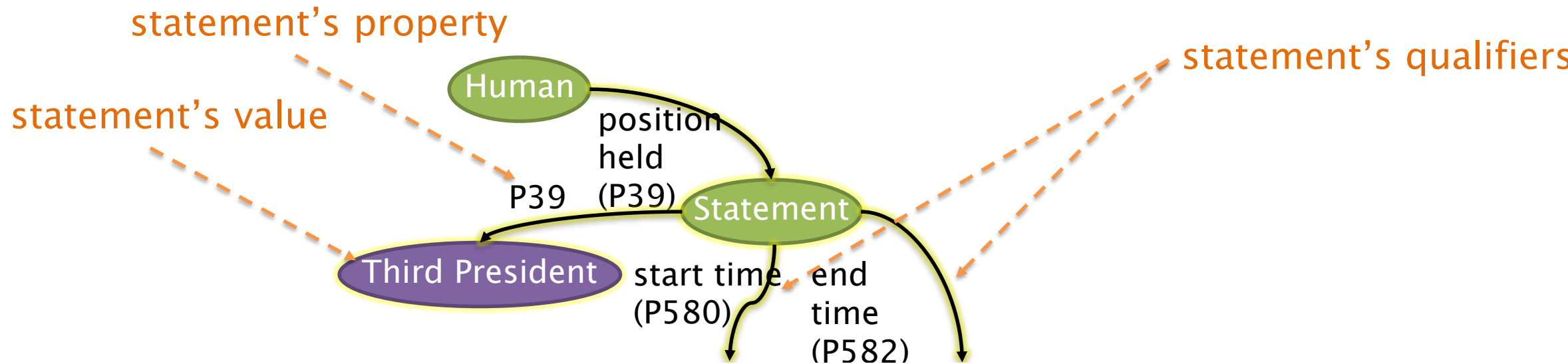


Third Presidents of National Council (Austria)

Source Modeling Problem



- Building semantic descriptions of tables
 - Describing data source using classes and properties in ontologies



Name	Entered Office	Left Office	Party
Thomas Prinzhorn	2002	2006	FPÖ
Eva Glawischnig-Piesczek	2006	2008	Grüne
Martin Graf	2008	2013	FPÖ

Third Presidents of National Council (Austria)

Main Idea



- Information of entities in KGs can help source modeling
⇒ need little training data

President of the National Council (Austria)

From Wikipedia, the free encyclopedia

List of third presidents [\[edit \]](#)

Name	Entered Office	Left Office	Party
Thomas Prinzhorn	2002	2006	FPÖ
Eva Glawischnig-Piesczek	2006	2008	Grüne
Martin Graf	2008	2013	FPÖ

Eva Glawischnig-Piesczek [\(Q93870\)](#)

Austrian politician

[edit](#)

member of political party

[Die Grünen](#)

position held

[Third President of the National Council of Austria](#)

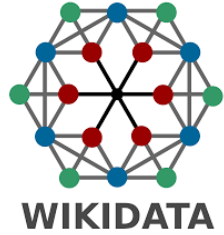
start time

30 October 2006

end time

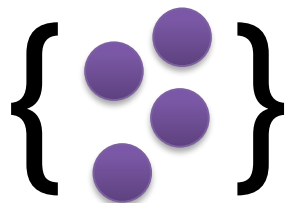
28 October 2008

Approach

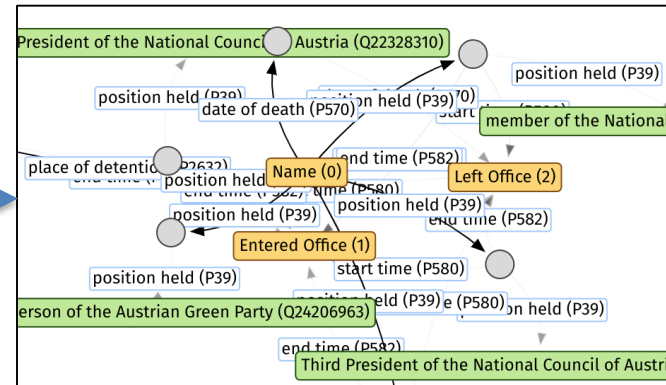


Name	Entered Office	Left Office	Party
Willi Braunerder	1996	1999	FPÖ
Thomas Prinzhorn	2002	2006	FPÖ
Eva Glawischnig-Piesczek	2006	2008	Grüne
Martin Graf	2008	2013	FPÖ

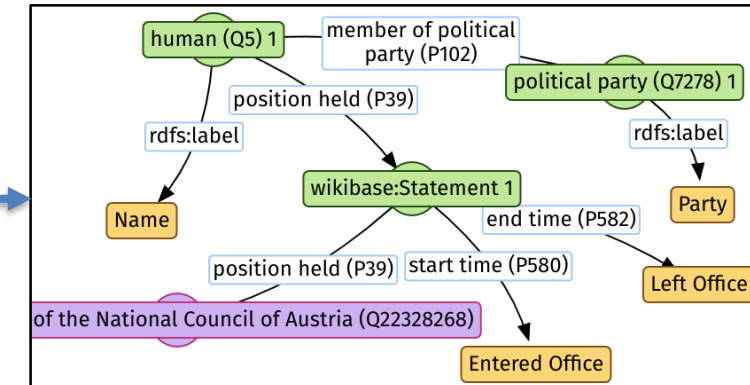
Linked table



Contextual values



Candidate Graph



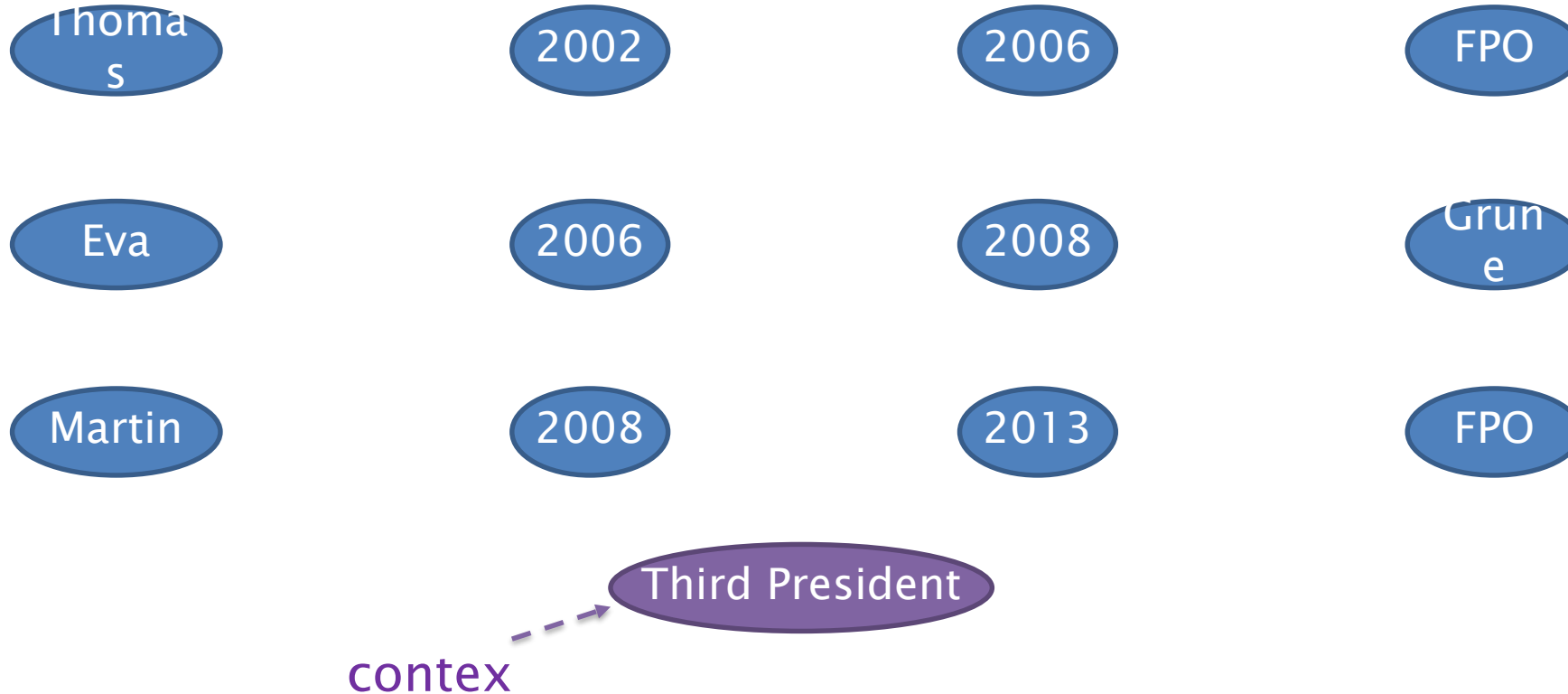
Semantic Description

Construct Candidate Graph: Discovering Links



- Create a graph of cells and co

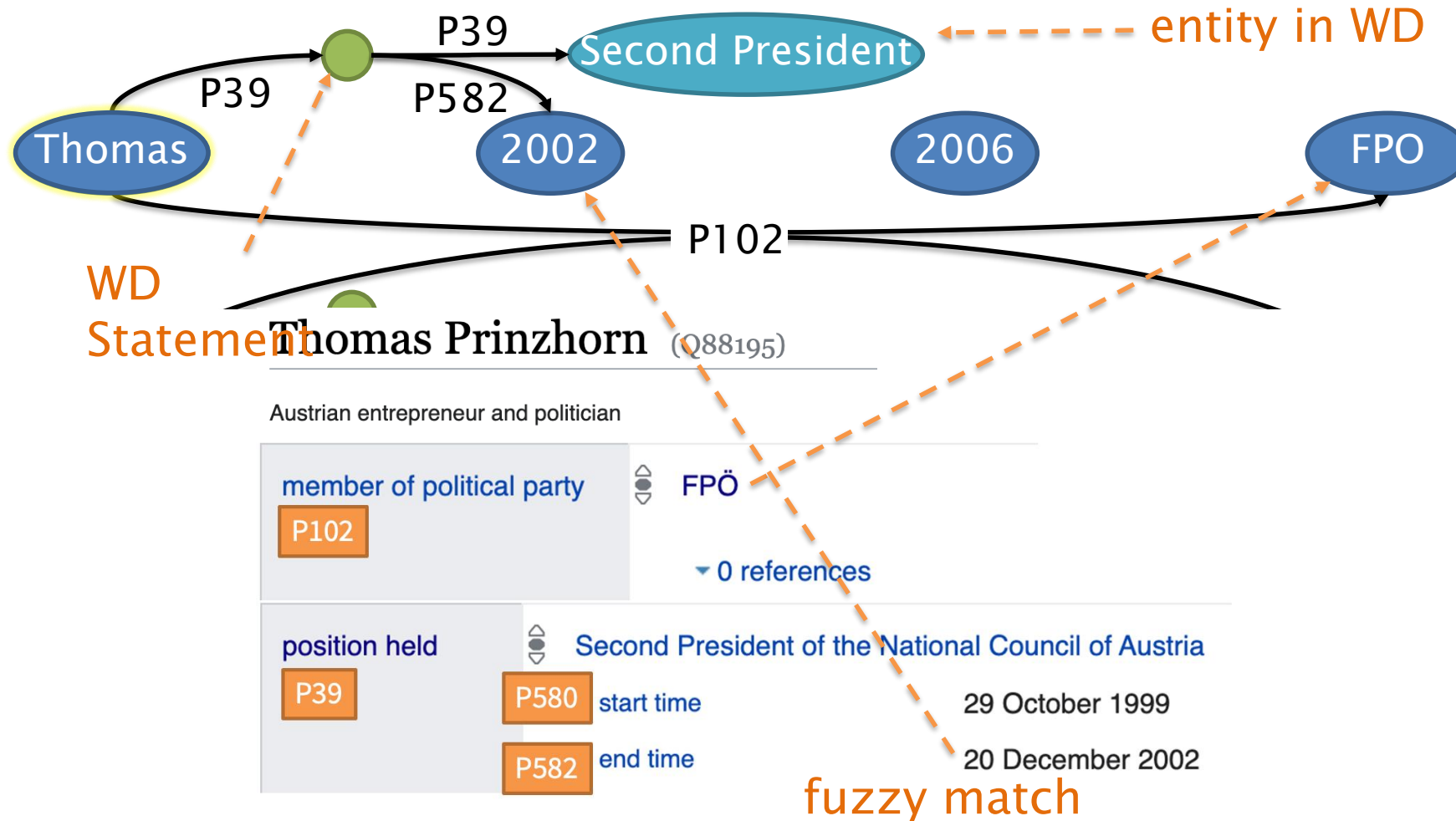
Name	Entered Office	Left Office	Party
Thomas Prinzhorn	2002	2006	FPÖ
Eva Glawischnig-Piesczek	2006	2008	Grüne
Martin Graf	2008	2013	FPÖ



Construct Candidate Graph: Discovering Links



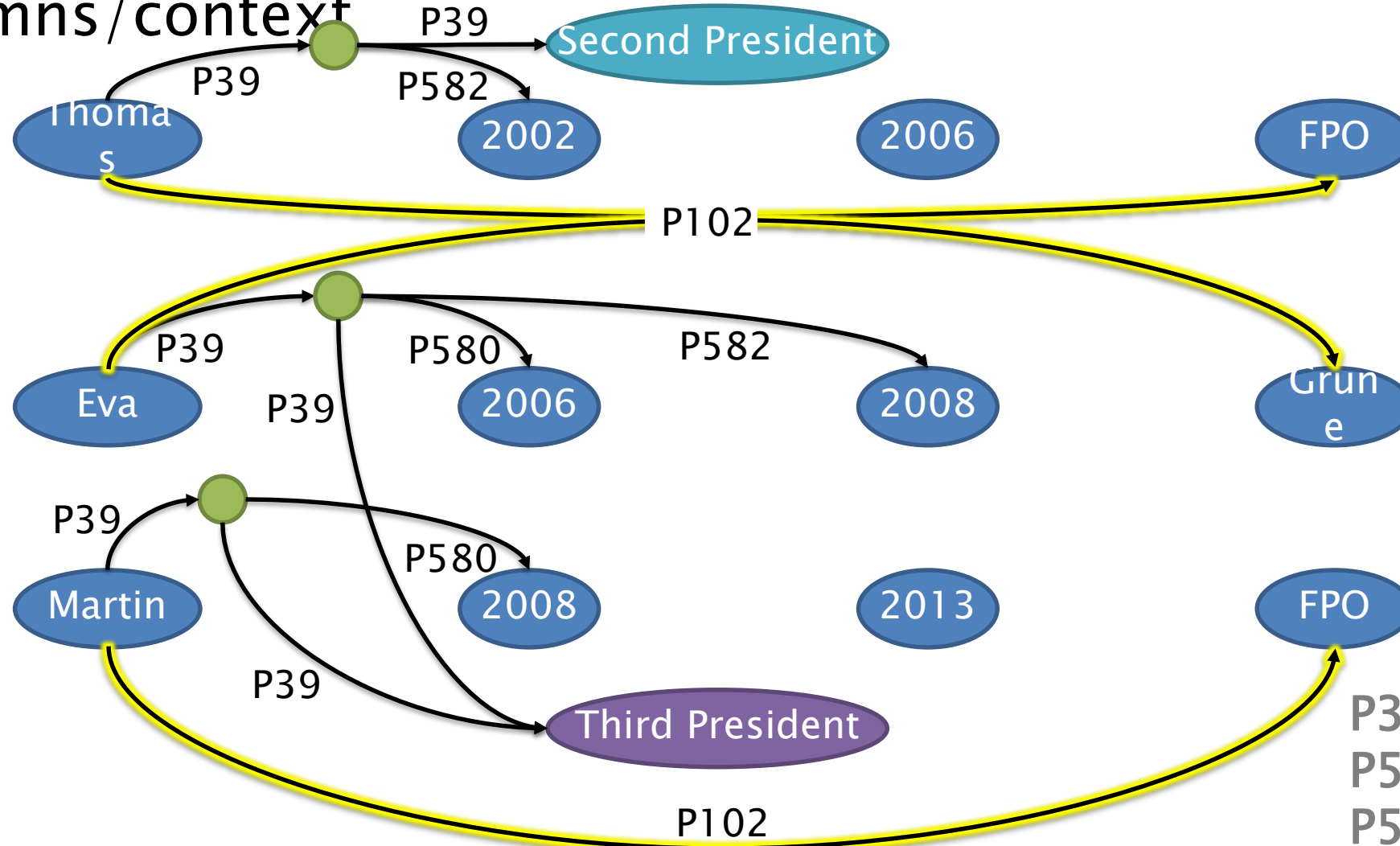
- Add links discovered from knowledge in Wikidata



Construct Candidate Graph: Summarization

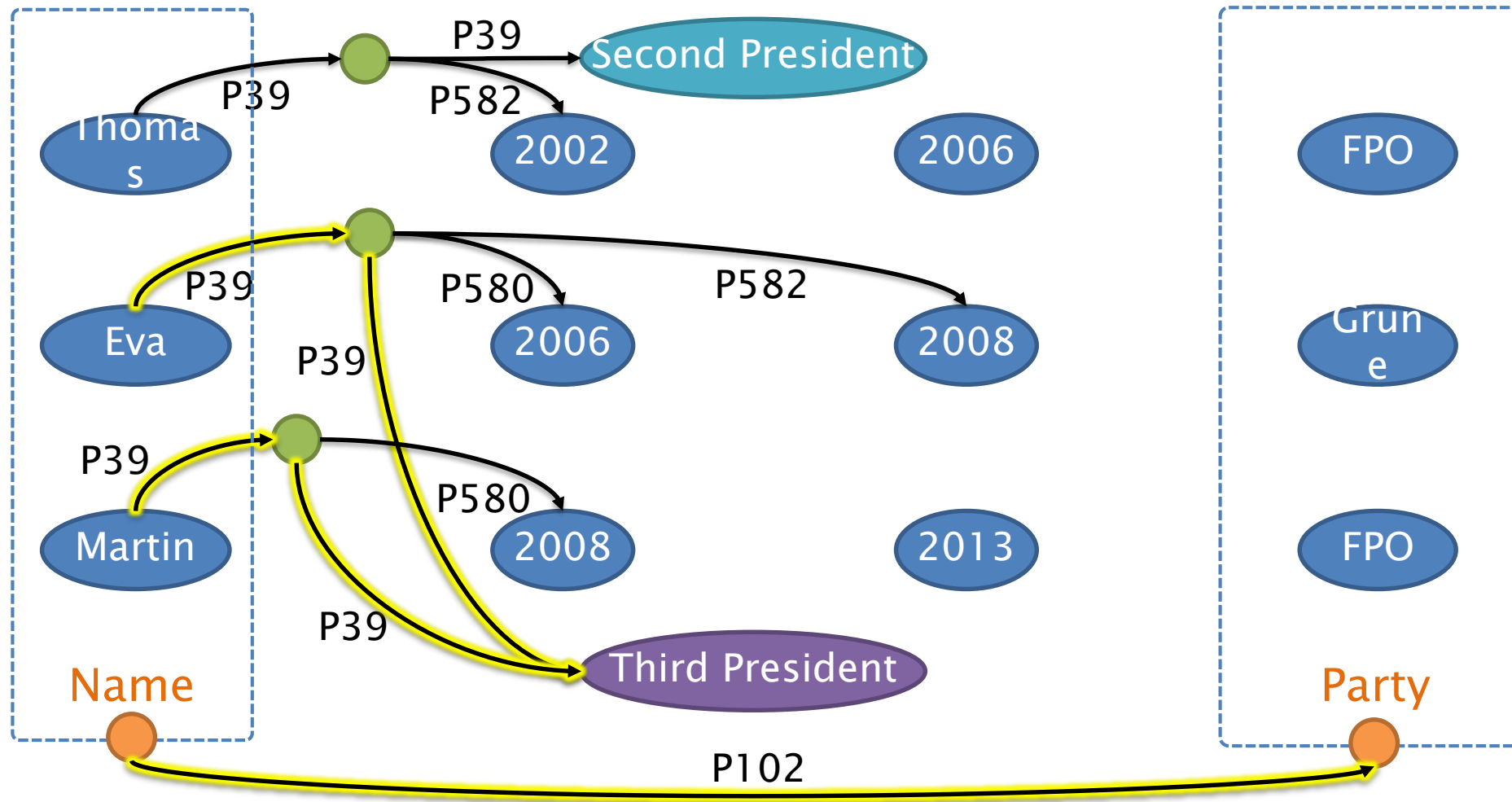


- Group links of cells from same source & target columns/context

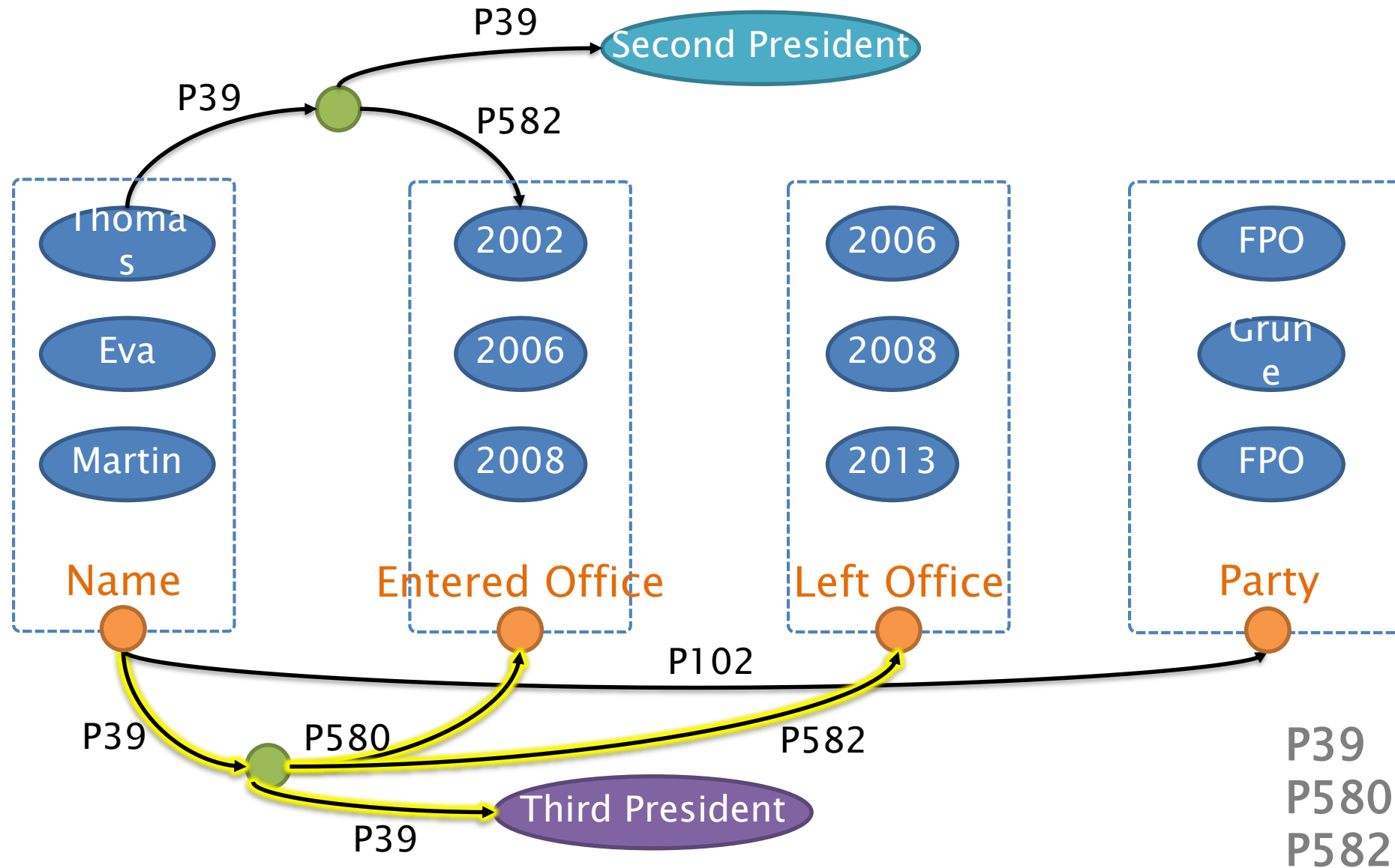


P39 : position held
P580: start time
P582: end time

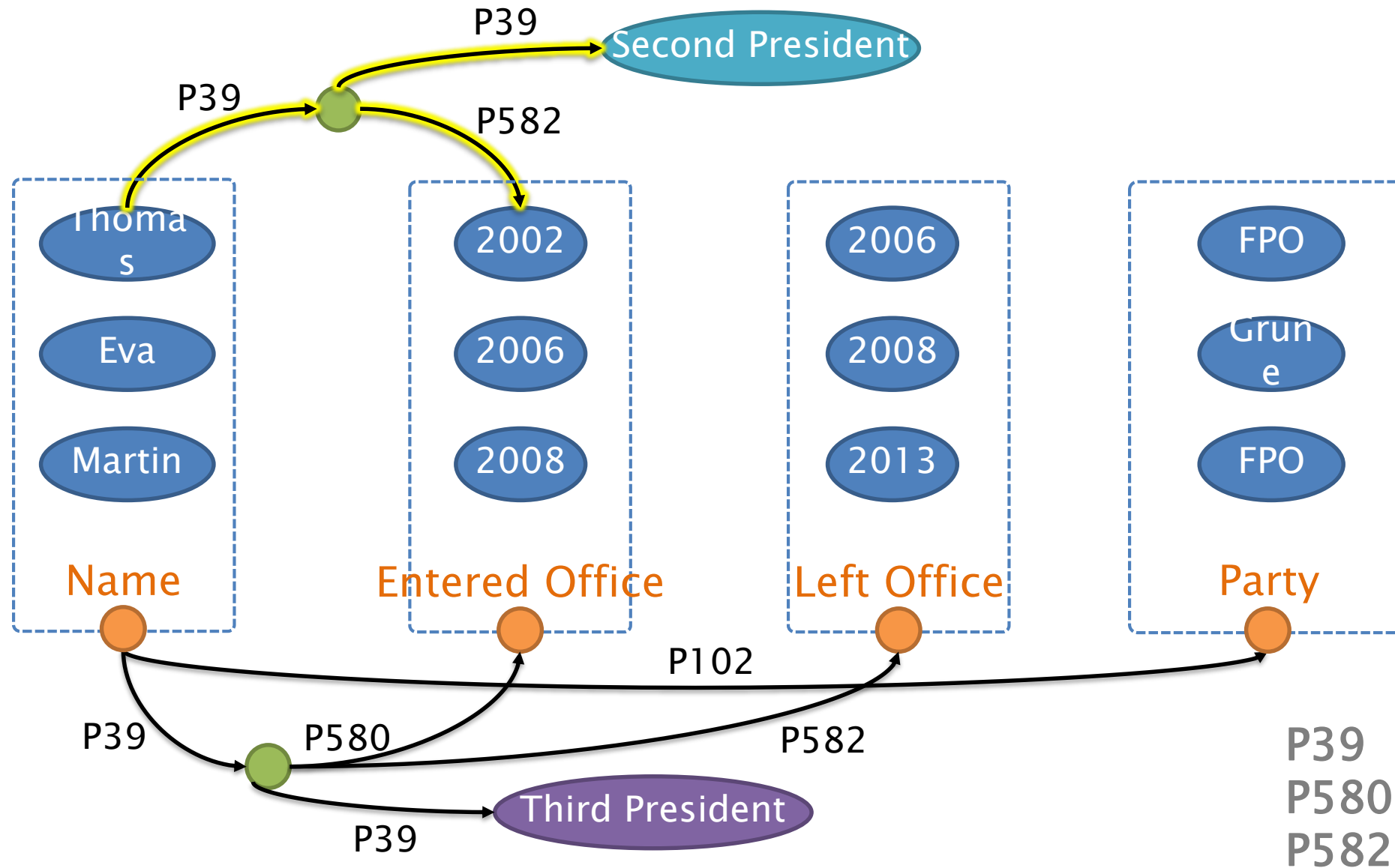
Construct Candidate Graph: Summarization



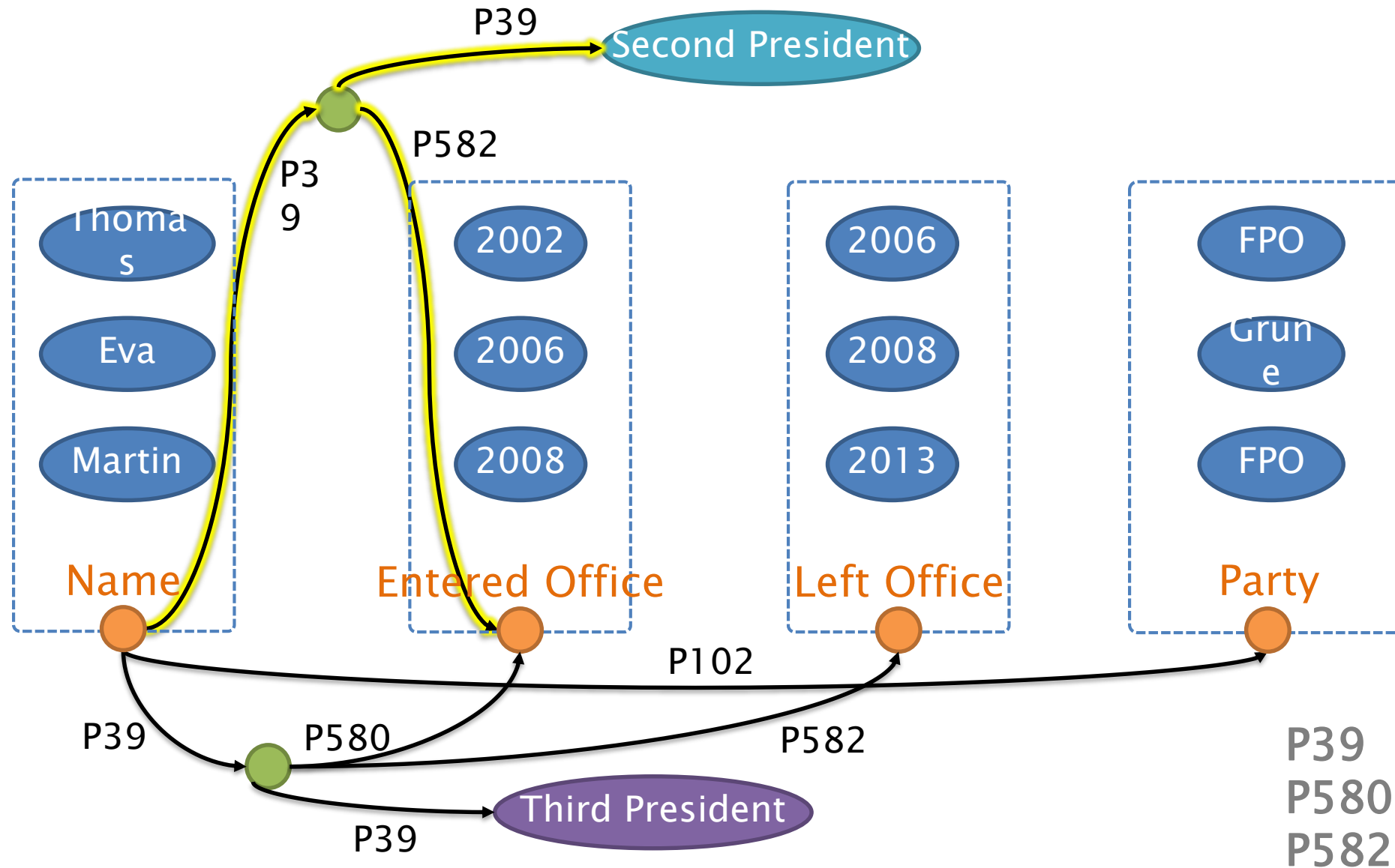
Construct Candidate Graph: Summarization



Construct Candidate Graph: Summarization



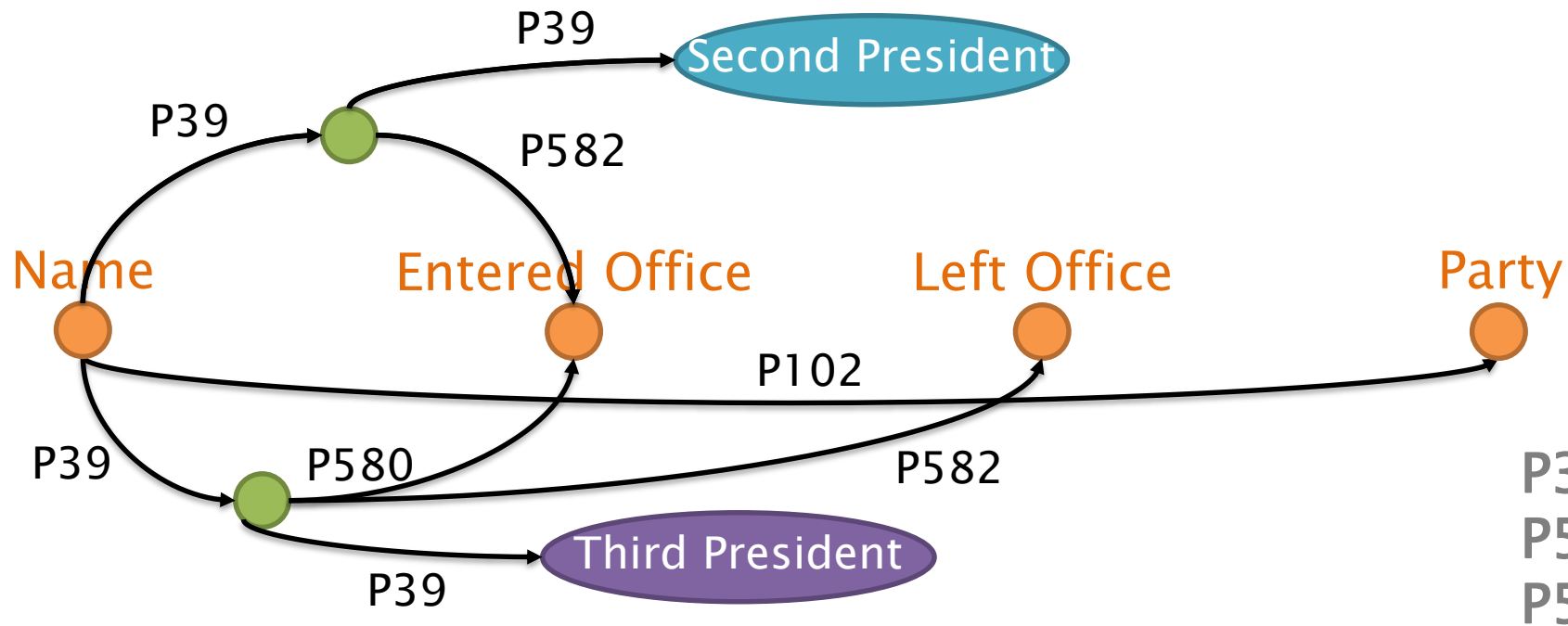
Construct Candidate Graph: Summarization



Construct Candidate Graph: Summarization



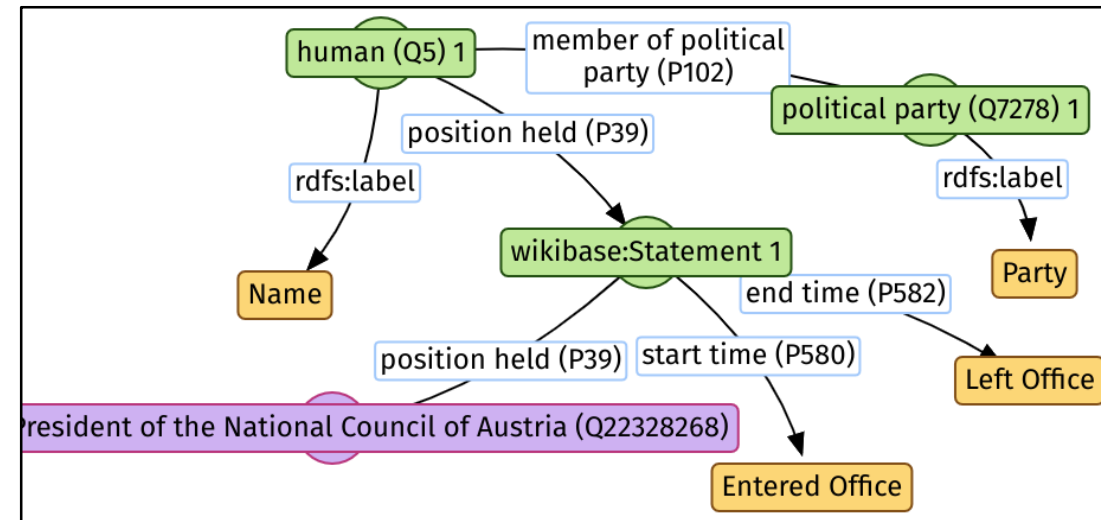
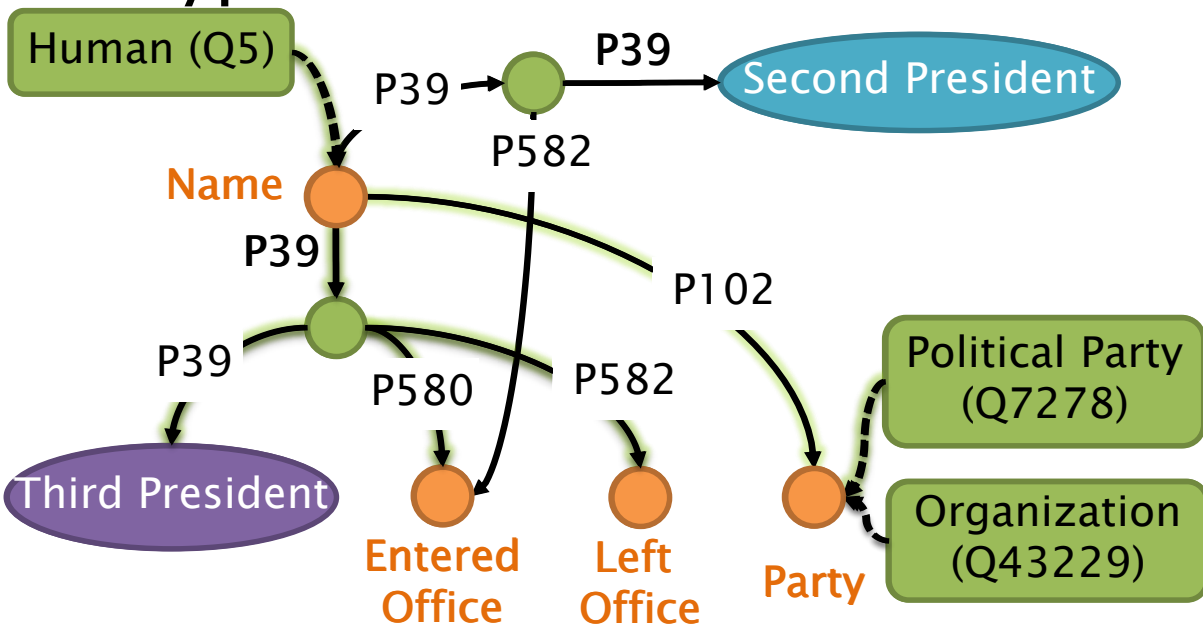
- Final candidate graph





After Building Candidate Graph

- Candidate (n-ary) relationships *from the candidate graph*
 - Candidate columns' types *from entities in table columns*
- ⇒ Need to select the most appropriate relationships and types.



Semantic Description

Candidate Graph

Approach



Inputs

- A target knowledge graph: Wikidata
- A linked relational table T
- A set of contextual values C

1. Construct candidate graph

2. Infer semantic description

Outputs:

- A semantic description of (T, C)

Collective Reasoning Problem



- Probabilistic Soft Logic (PSL)

“A probabilistic graphical models framework using first-order logic”

- Two main elements: **predicates** and **rules**

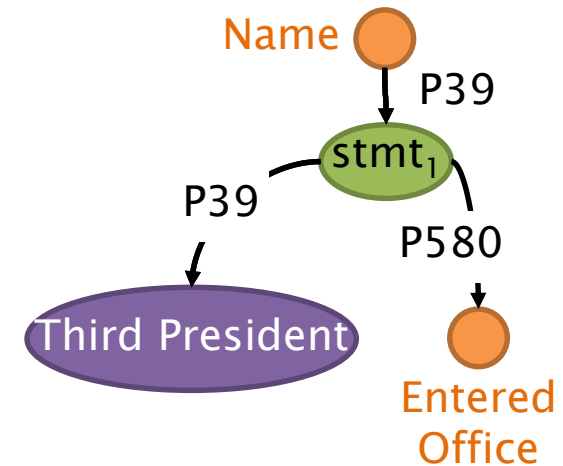
- Predicates have “soft” value in $[0, 1]$
- Rules converted to exponential function to approximate $P(\mathbf{x})$

PSL Predicates (examples)



- **CorrectRel(N_1 , N_2 , P)**: if a relationship is correct

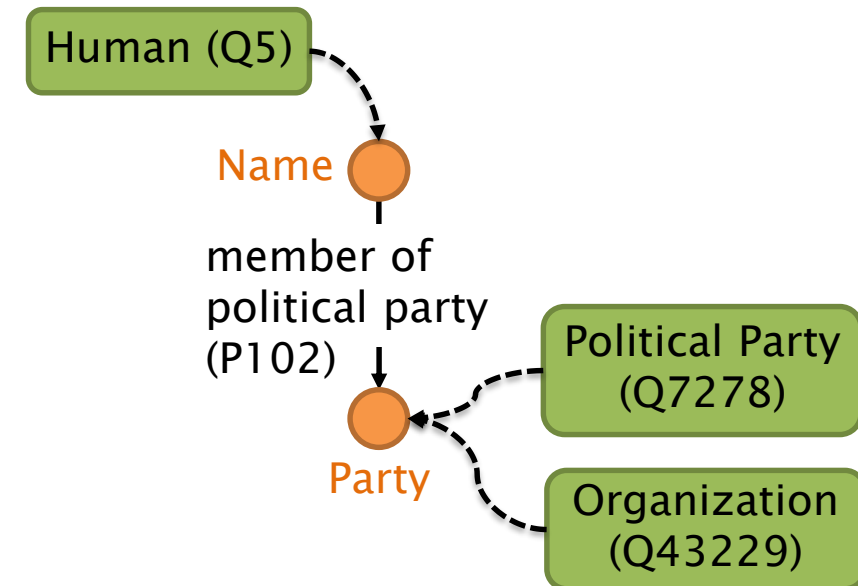
- CorrectRel(Name, stmt₁, P39)
- CorrectRel(stmt₁, Entered Office, P580)
- CorrectRel(stmt₁, Third President, P39)



- **CorrectType(N_1 , T)**: if a column type assignment is correct

- CorrectType(Party, Organization)
- CorrectType(Party, Political Party)
- CorrectType(Name, Human)

- ... and more



P39: position held P580: start time P582: end time



PSL Rules (examples)

1. By default, relationships/types are incorrect

1 a. $\neg \text{CorrectRel}(N_1, N_2, P)$

1 b. $\neg \text{CorrectType}(N_1, T)$

2. Relationships/types are correct/incorrect based on evidence

2a. $\text{FreqMatch}(N_1, N_2, P) \rightarrow \text{CorrectRel}(N_1, N_2, P)$

2b. $\text{FreqDiff}(N_1, N_2, P) \rightarrow \neg \text{CorrectRel}(N_1, N_2, P)$

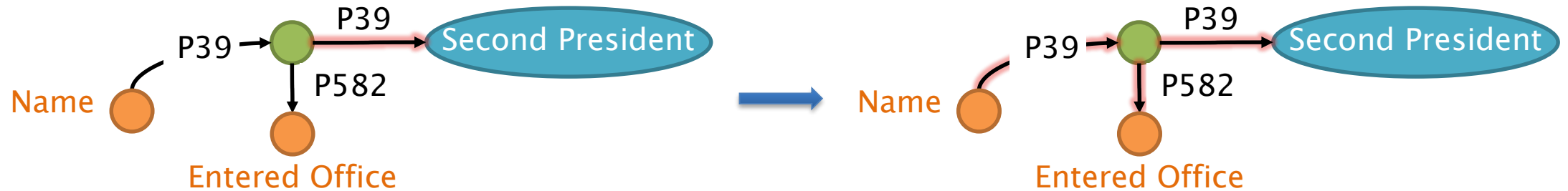
2c. $\text{FreqTypeMatch}(N_1, T) \rightarrow \text{CorrectType}(N_1, T)$

2d. ...and more

PSL Rules (examples)

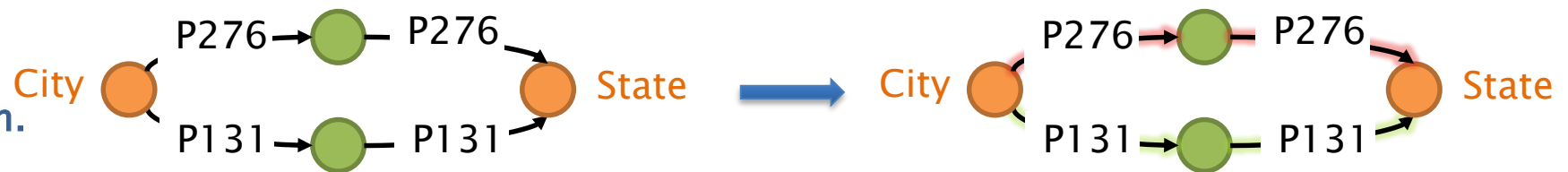


3. If a statement value is incorrect, then the statement's qualifiers are also incorrect



4. We prefer fine-grain properties than high-level properties

location (P276)
is parent of
located in admin.
area (P131)

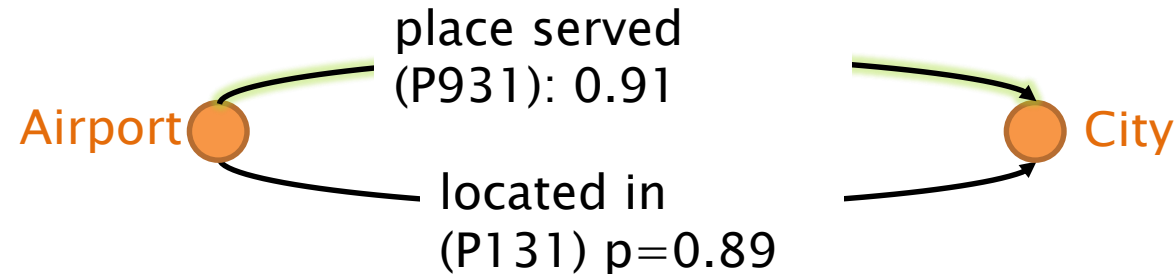


5. ...and more

Post-Processing



- PSL outputs probability of each relationships and types.



- Use BANK algorithm to choose the most probable relationships
 - Avoid unnecessary loops
 - Prefer tree structure if possible



Evaluation of GRAMS



- Collective reasoning is beneficial
 - Avoid cascading errors from subject column detection phase
 - Handle complex schema: multiple entities' types and n-ary relationships

		Dataset	Method	CPA			CTA		
				Precision	Recall	F ₁	Precision	Recall	F ₁
Wikipedia Tables	250WT	MantisTable	0.535	0.442	0.484	0.928	0.331	0.488	
		MantisTable*	0.559	0.569	0.564	0.940	0.394	0.556	
		BBW	0.796	0.123	0.214	0.850	0.233	0.367	
		BBW*	0.740	0.559	0.638	0.759	0.777	0.768	
		GRAMS-ST	0.526	0.681	0.594	-	-	-	
		GRAMS	0.824	0.650	0.726	0.819	0.813	0.816	
Synthetic Tables	SemTab2020	MantisTable	0.985	0.976	0.981	0.977	0.800	0.880	
		BBW	0.996	0.995	0.995	0.980	0.980	0.980	
		GRAMS-ST	0.990	0.989	0.990	-	-	-	
		GRAMS	0.996	0.994	0.995	0.982	0.981	0.982	

MantisTable* and BBW* are modified to retrieve correct subject

Related Work

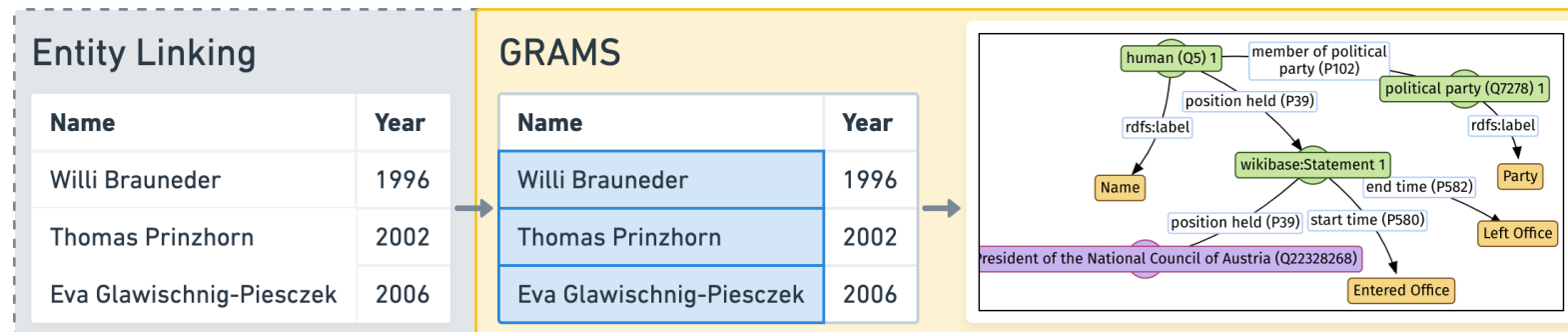


	Method		Data Hungry	Modeling Capabilities		
				Handle Literal Columns	Handle Qualifiers	Denormalized Tables
Custom Ontologies	Taheriyani et al. 2016		Y	Y	Y	Y
	Vu et al. 2019		Y	Y	Y	Y
KG Ontologies	Iterative Method	Ritze et al. 2015	–	Y	N	N
		Zhang et al. 2017	–	Y	N	N
		SemTab systems	–	Y	N	N
	Graphical Models	Limaye et al. 2010	–	N	N	Y
		Mulward et al. 2013	–	N	N	Y
		GRAMS	–	Y	Y	Y

Discussion and Future work



- **Contribution:** A novel graph-based approach, GRAMS, for building semantic descriptions of Wikipedia Tables.
 - The candidate graph makes it easy to represent and discover n-ary relationships.
 - Using PSL to collectively infer correct relationships and types.
- **Future work:**
 - Handle unlinked tables



- Generate large labeled dataset from Wikipedia tables to train semantic modeling systems