

# A Role of Ontology in Social Data Analytics

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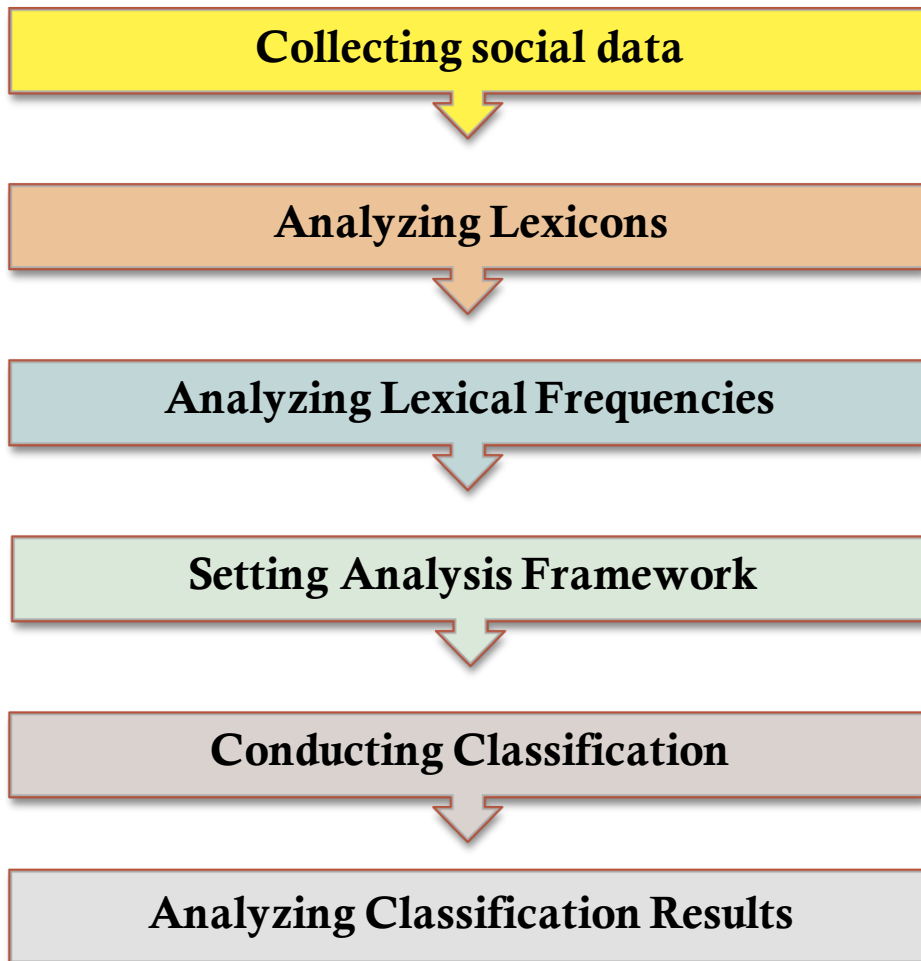
# Talk Outline

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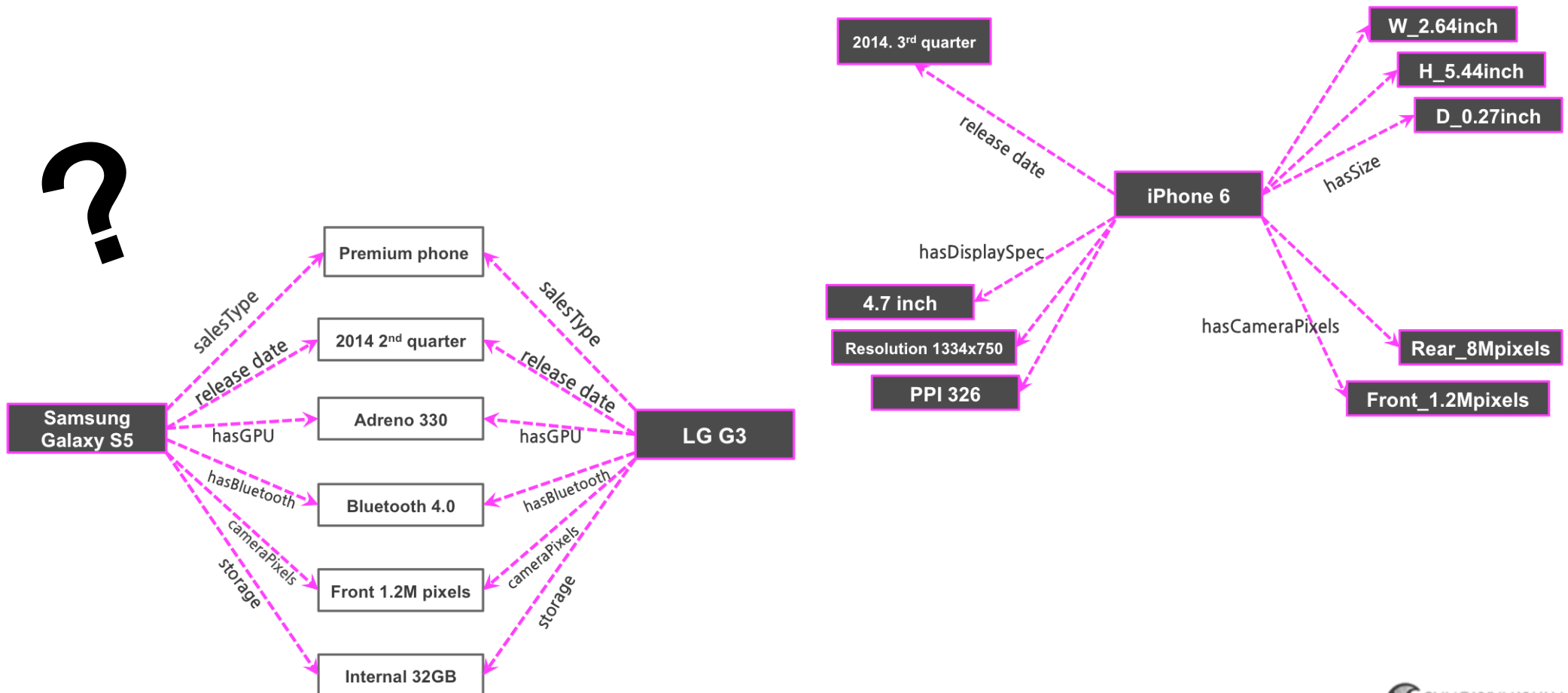
1. Research Background
2. Research Purpose
3. Research Process
4. Research Findings
5. Further Study

- Vocabulary Management for Social Data Analytics
  - Tends to be managed in a solo
  - Lack of reusability and interoperability from one app to the other
  - Speedy vocabulary management is needed for live data analytics
- Much efforts and time spent in collecting terms and devising a structure to them
  - Requires a considerable time to establish a VOC management system for a new project
- Relationships among terms and additional details on each term need to be established
  - Useful in deriving insights by taking the relationships into account

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- To understand what roles ontology can play in social data analytics



3.1 Constructing 'smartphone' ontology

3.2 Ingesting DBPedia data related to smartphones

3.3 Visualizing 'smartphone' ontology

3.4 Ontology Benefits in Social Data Analytics (SDA)

- Limitations of Current SDA Systems
- Improvement of Current SDA Systems

# 3.1 Constructing a 'Smartphone' Ontology

Managing smartphone ontology with the following SKOS and OWL vocabularies:  
 skos:Concept, skos:broader, skos:narrower, skos:prefLabel, skos:altLabel

<LG Vocabulary Scheme>

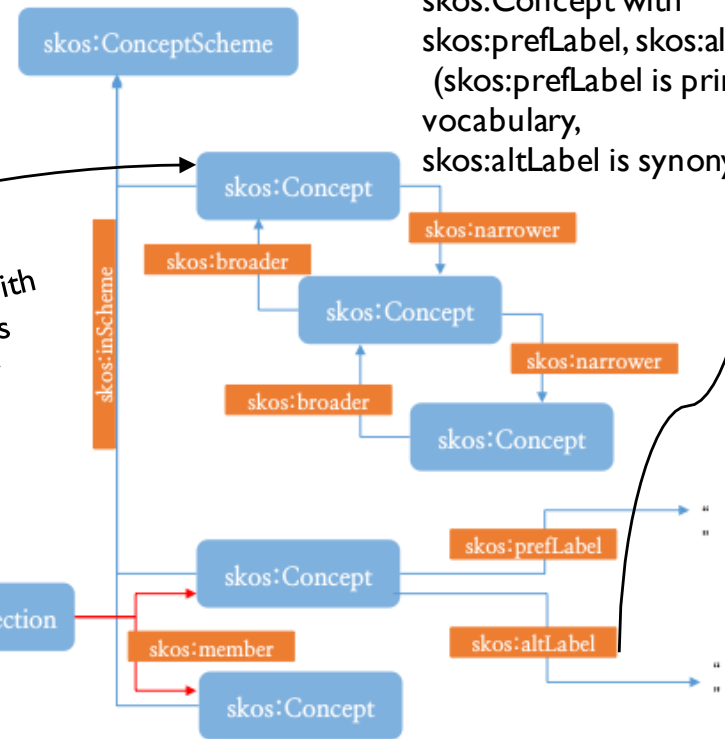
L1	L2	L3	L4	L5
LG Electronics	Smart Phone	Optimus	general	OptimusLTE 2
				OptimusLTE 3
		G Series	premium	G3
				G2
				G PRO
Nexus	general	Nexus 4		
Samsung Electronics	Smart Phone	Galaxy	general	Galaxy Grand
				Galaxy POP
			premium	Galaxy Note3
				Galaxy S4
				Galaxy S5
Apple	Smart Phone	iPhone	premium	iPhone5
				iPhone5s
				iPhone6
			general	iPhone6plus
				iPhone5c

<SKOS Basic Structure>

Managing Smartphone Title with  
 skos:Concept to set relations  
 skos:broader, skos:narrower

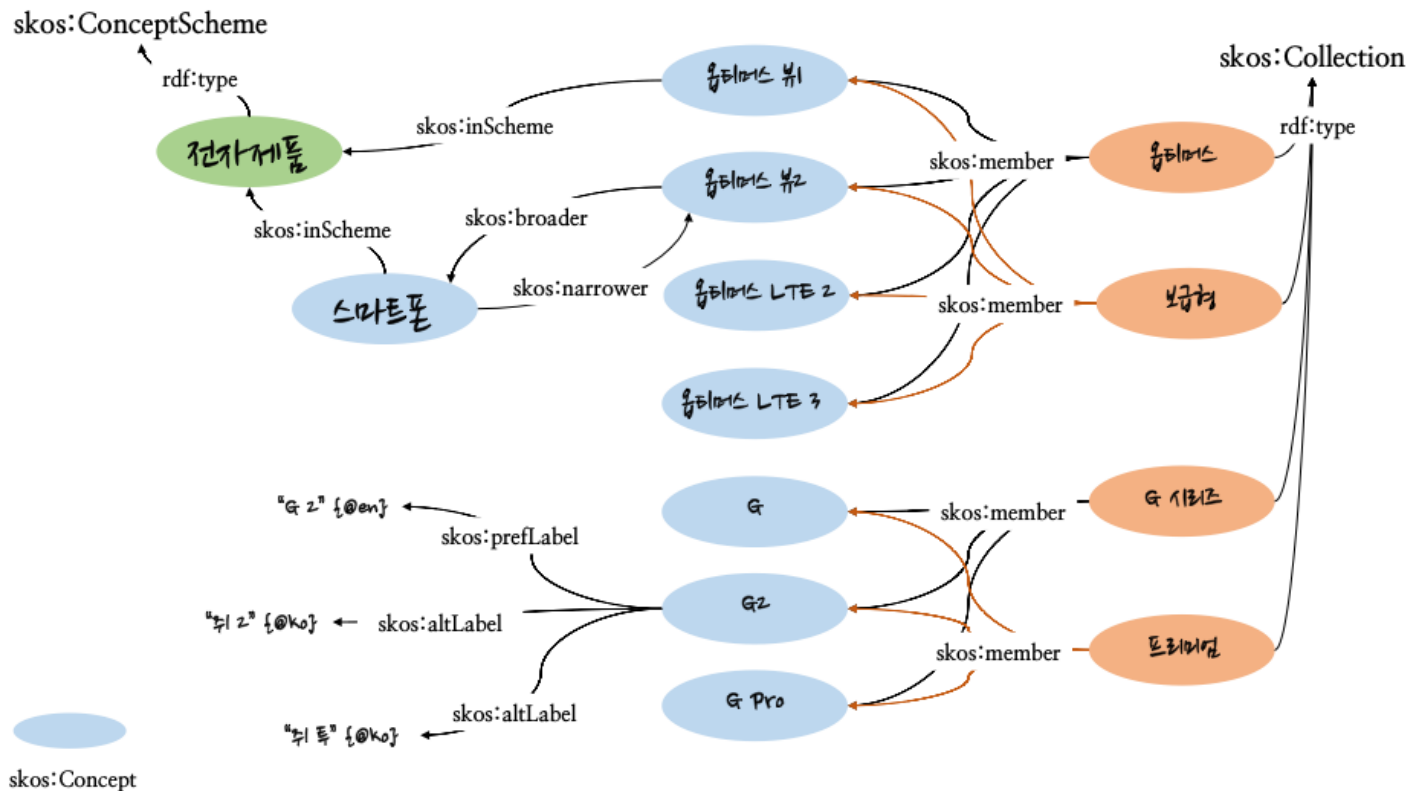
Providing group  
 information using  
 skos:Collection

Providing synonym for  
 skos:Concept with  
 skos:prefLabel, skos:altLabel  
 (skos:prefLabel is primary  
 vocabulary,  
 skos:altLabel is synonym)



# 3.1 Constructing a 'Smartphone' Ontology

## <SKOS/OWL-Based Vocabulary Management >



### SKOS/OWL-based Management

- **SKOS/OWL-based vocabulary management allows LG-CNS to import external data and provides interoperability.**



## Ingesting relevant data (firms and personal info) into ontology

### DBpedia

- As of August 2014, 500 datasets. The center circle is Dbpedia.
- LOD De facto, Huge amount of data
- 191 Korean firm information out of 50,000 firm information.
- 200 Korean out of 1,440,000 personal information

### Querying Data

Extract metadata for massive data extracted from Dbpedia

### Downloading Data

SPARQL Endpoint – save as spreadsheet

### Saving as a spreadsheet

Have a program to import spreadsheet data into ontology

# 3.3 Visualizing 'Smartphone' Ontology

## TM Analysis Expanded

- Protégé PlugIn, Relfinder, and D3 provide different views of data

## AS-IS

- Proximity measured by word-frequency
- Raw data must be consulted to understand relationships among terms



AS-IS TM

## TO-BE

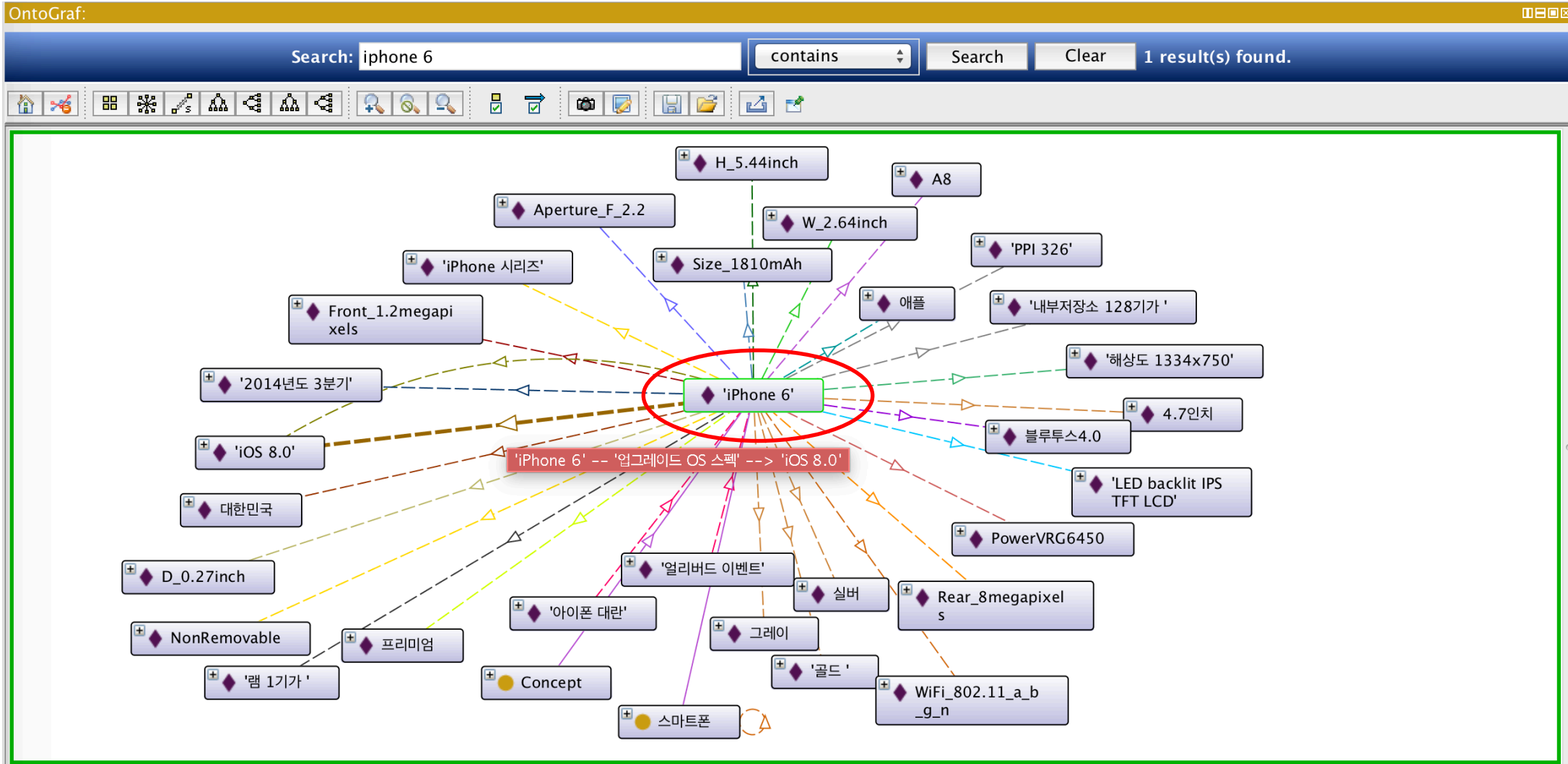
- Enhanced views of data
- Relationships among keywords are provided



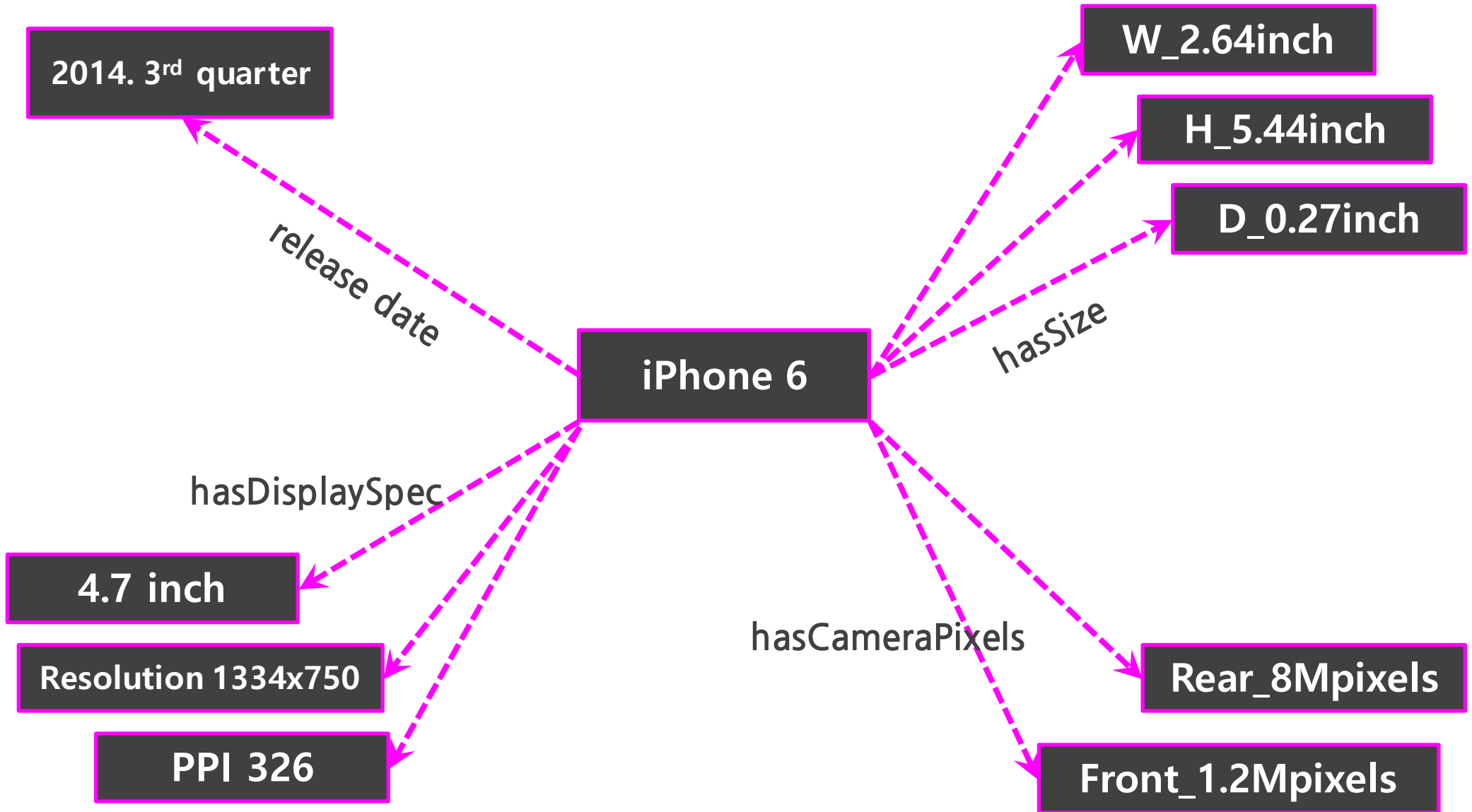
# 3.3 Visualizing 'Smartphone' Ontology: Type A

## Main features of Type A

- Providing detailed info on each term



Mouse-over will display relationship



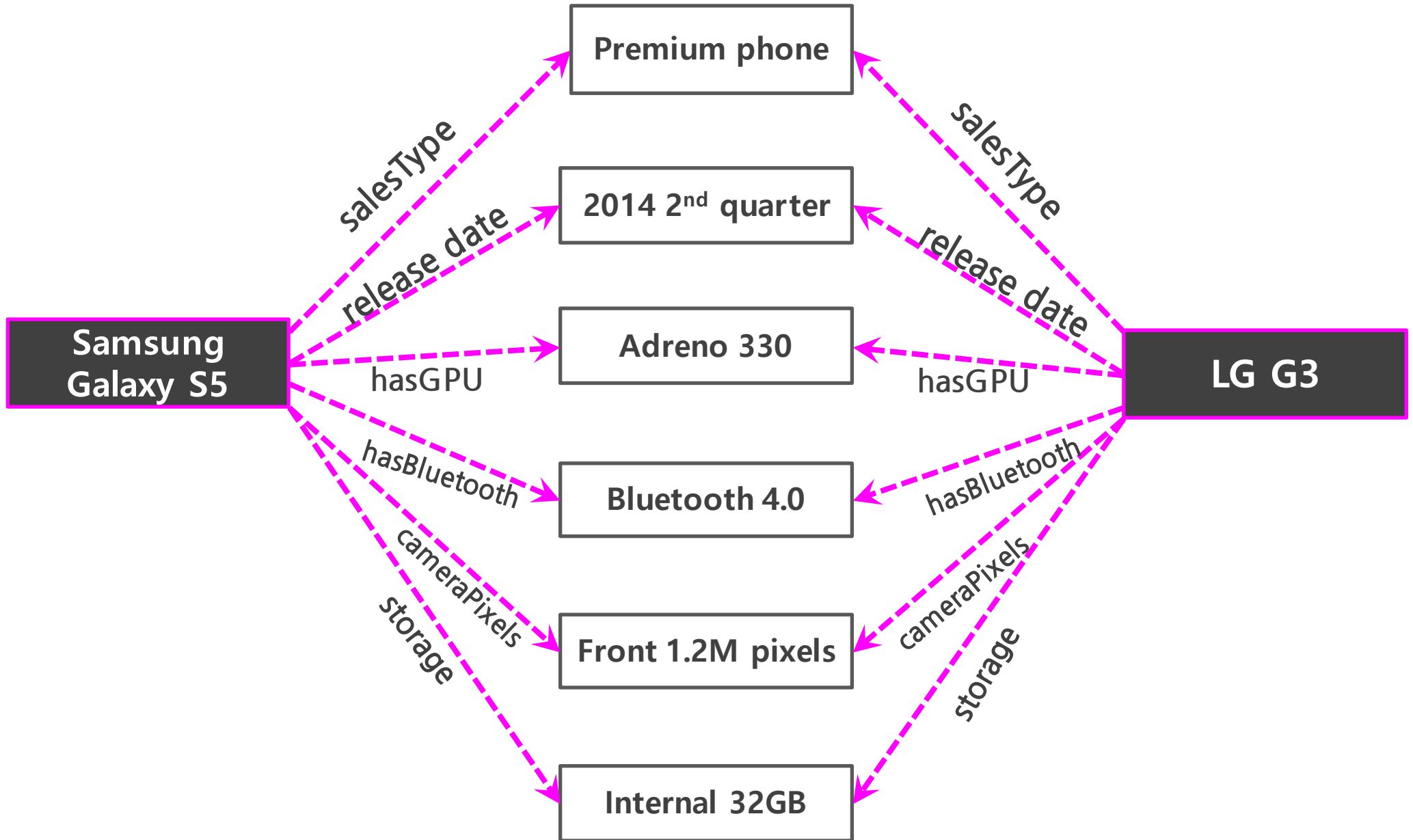
## Main features of Type B

- When you want to know relationships between two terms

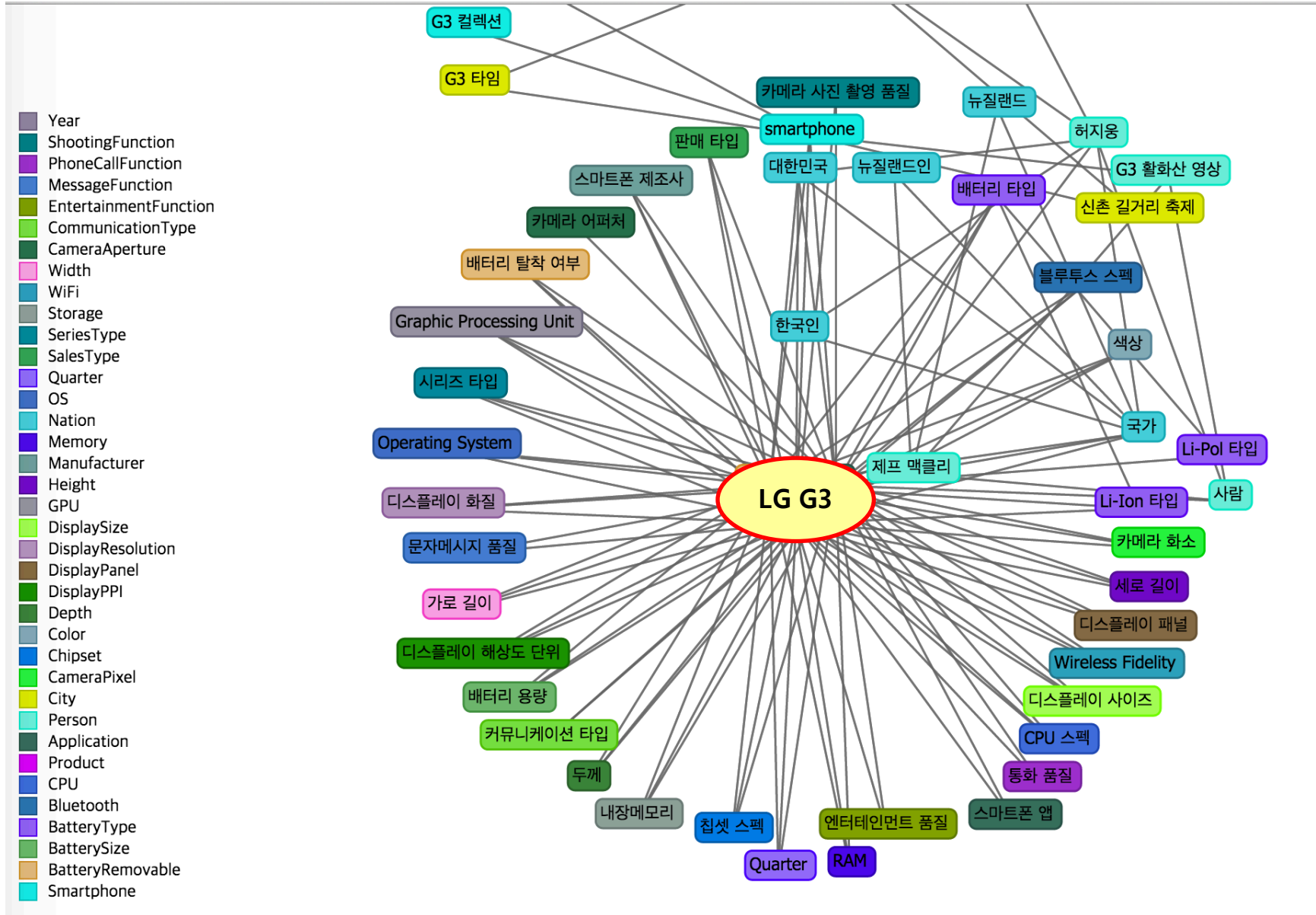
The screenshot shows the RelFinder interface on the left and a large ontology graph on the right. The interface includes a search area with 'G3' and 'S5' entered, a filter table, and a language dropdown set to 'en'. The ontology graph shows two terms circled in red: 'lg 옵티머스 G3' and '스파이브'. They are connected to various other terms through relationships like 'type', 'hasGPU', 'madeYearQu...', 'hasSalesType', 'manufacture...', 'producedBy', and 'advertisedBy'. Other terms include 'Smartphone', 'Adreno 330', 'SecondQuarterOf2014', '고급형 폰', 'SouthKorea', 'Kim Soo Hyun', 'Jeon Ji Hyun', 'Samsung electronics', 'LGElectronics', 'Heo Ji-Woong', 'SouthKorea', 'bornInNation', 'manufacture...', 'producedBy', and 'advertisedBy'. The graph also shows relationships to 'Concept', 'Smartphone', 'Adreno 330', 'SecondQuarterOf2014', '고급형 폰', 'SouthKorea', 'Kim Soo Hyun', 'Jeon Ji Hyun', 'Samsung electronics', 'LGElectronics', 'Heo Ji-Woong', 'SouthKorea', 'bornInNation', 'manufacture...', 'producedBy', and 'advertisedBy'.

number of objects	num	vi
1	10/10	
2	5/5	

- For example, what are the common features of G3 and S5?



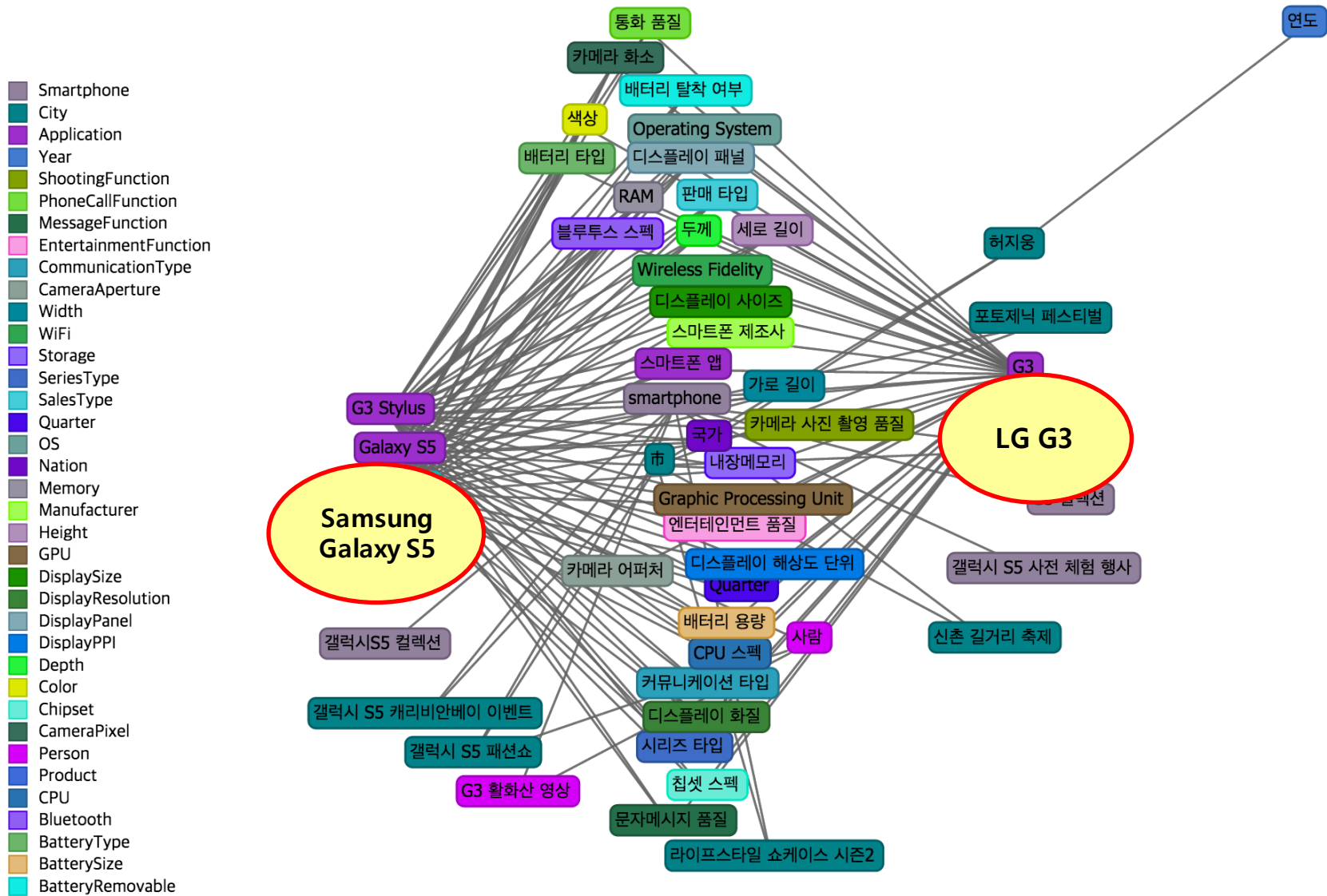
## Combining Type A and B



Search Results of 'LG G3'



# 3.3 Visualizing 'Smartphone' Ontology: Type C



**Search Results of 'LG G3' and 'Samsung Galaxy S5'**



To identify the benefits of ontology in social data analytics

- Conducted Delphi surveys involving experts in social data analytics and social marketing
- Extracted criteria to assess the benefits of the ontology in social data analytics

### Experts in Social Data Analytics

- ✓ Those who are savvy in setting up rules and stop words to take advantage of a particular domain analysis
- ✓ Those who are savvy in building a comprehensive service such as a TM dashboard
- ✓ Those who have worked for more than 2 years in the job mentioned above

### Experts in Social Marketing

- ✓ Those who have expertise in analyzing markets and developing an effective marketing plan utilizing SDA tools
- ✓ Those who have experience in writing 'insight report' on market and product analysis

### I<sup>st</sup> Delphi Study

#### Questions used

- What are the benefits of the current SAS Text Mining (TM) tools in conducting SDA?
- What are the shortcomings of the current SAS TM tools in conducting SDA?
- To have a more effective SAS TM tools in SDA, what functions should be improved

### Quantitative Evaluation

### Questions used for study

- Was it helpful to use ontology in deriving insights?
- In understanding keyword relationships
- In obtaining further details on each term
- In constructing a comprehensive system for social data analytics
- In combining rule-based analysis and machine learning
- In getting contextual meaning of each term in a particular domain
- In obtaining up-to-date information

**Ontology  
Assessment Criteria  
extracted from survey**

**Experts in Social  
Data Analytics**



Insights  
Keyword relationships  
Additional information  
Comprehensive system  
Innovative analysis  
Contextual meaning  
Up-to-date Information



**Experts in Social  
Marketing**

### Results of Quantitative Evaluation: Experts in Social Data Analytics

- **7 Point Scale**

Variable	Avg	STD	Normal	T-test p-v	Wilcoxon p-v
Insights	5.286	0.731	0.712	0.341	0.303
Keyword relationships	5.286	0.591	0.744	0.248	0.221
Additional info	5.048	0.591	0.450	0.838	0.785
Comprehensive system	5.333	1.054	0.952	0.435	0.336
Innovative analyses	5.238	1.384	0.998	0.665	0.598
Contextual meaning	5.333	0.923	0.910	0.376	0.336
Up-to-date info	4.476	1.425	0.866	0.368	0.380

### Results of Quantitative Evaluation: Experts in Social Marketing

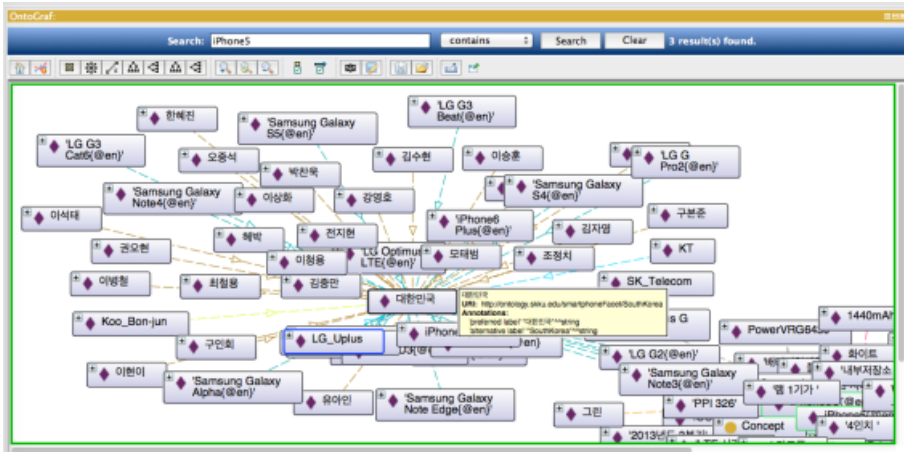
- **7 Point Scale**

Variable	Avg	STD	Normal	T-test p-v	Wilcoxon p-v
Insights	5.067	0.723	0.925	0.847	0.892
Keyword relationships	5.067	0.760	0.949	0.854	0.705
Additional info	5.533	0.606	0.955	0.120	0.109
Comprehensive system	5.333	0.408	0.759	0.142	0.102
Innovative analyses	4.733	1.140	0.993	0.629	0.715
Contextual meaning	4.800	0.960	0.990	0.666	0.581
Up-to-date info	4.476	1.193	0.908	0.374	0.336

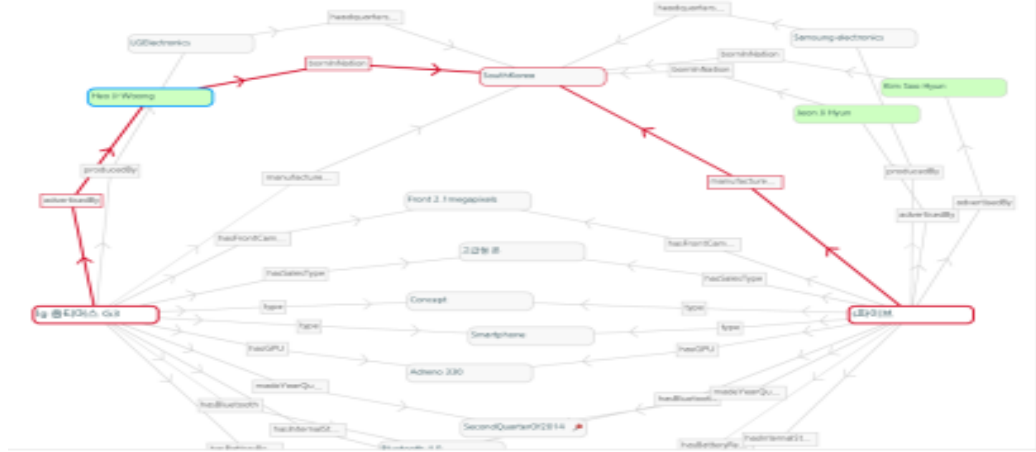
## Qualitative Evaluation I

- We asked why they prefer a particular visualization over the others per question (only those Qs that demonstrated significant difference)?
- Experts in Social Data Analytics
  - Insights, Keyword relationships, Innovative analyses (3)
- Experts in Social Marketing
  - Insights, Comprehensive systems (2)

### Type A: Additional Info



### Type B: Relationship





### Qualitative Evaluation 2

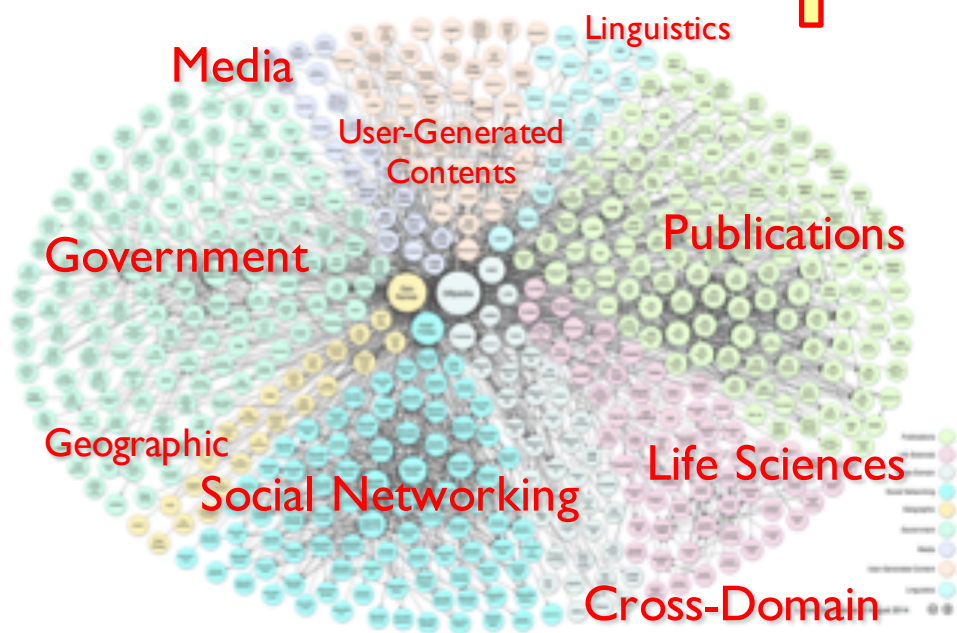
- Further analysis on deriving insights
  - Experts in Social Data Analytics: Preferred Type B
    - Readability, Identifying Relationships of each term
  - Experts in Social Marketing: Preferred Type A
    - Additional detail info was more important to them in gaining insights
    - Identifying relationships was optional, not crucial to them

### Qualitative Evaluation 3

**What are additional roles that ontology can play in SDA?**

#### Additional Roles of Ontology

- Experts in Social Market did not submit any idea on this.
- Experts in Social Data Analytics suggested that the following features are desired from the beginning of ontology setup.
  - Definition
  - Synonyms
  - Relationships



Datasets by topical domain.

Topic	Datasets	%
<a href="#">Government</a>	183	18.05%
<a href="#">Publications</a>	96	9.47%
<a href="#">Life sciences</a>	83	8.19%
<a href="#">User-generated content</a>	48	4.73%
<a href="#">Cross-domain</a>	41	4.04%
<a href="#">Media</a>	22	2.17%
<a href="#">Geographic</a>	21	2.07%
<a href="#">Social web</a>	520	51.28%
<b>Total</b>	<b>1014</b>	

- Global ID Management
- Ways to integrate LOD into ontology to enhance SDA

**Thank you for your attention!!  
Any questions?**