

Fully Exploit Hierarchical Structure for Self-Supervised Taxonomy Expansion

By Suyuchen Wang, Ruihui Zhao, Xi Chen, Yefeng Zheng, and Bang Liu

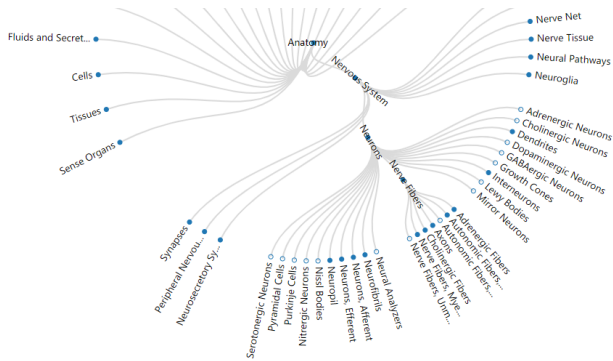
To Appear in The Web Conference 2021 (WWW'21)

2021.02.27

By 王苏羽晨



Amazon Product Taxonomy



Medical Subject Headings(MeSH)

Definition

- Taxonomy: A hierarchical structure modeling “is-A” or **hypernymy** relations

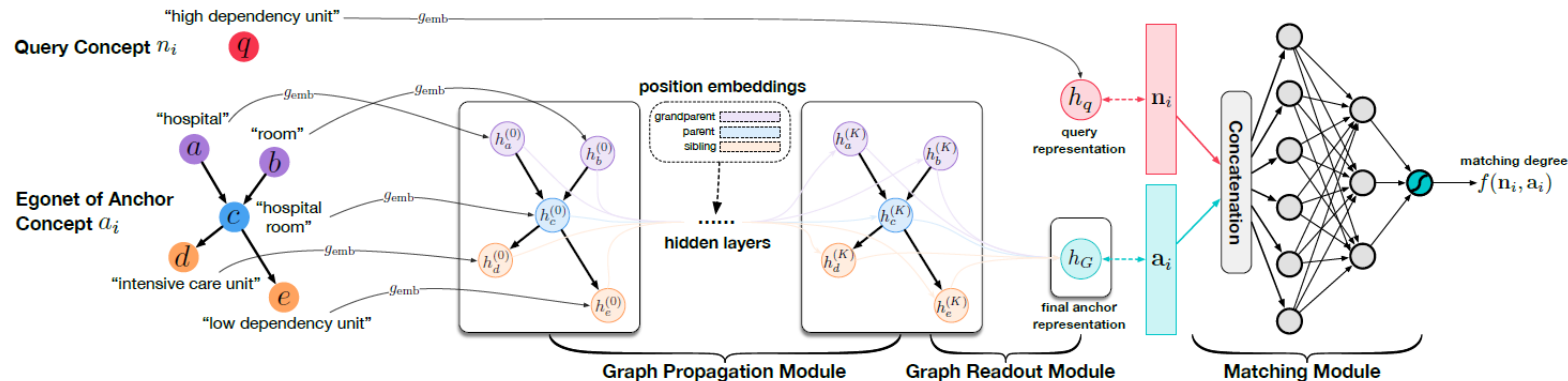
Problem

- Taxonomy curated by domain experts is **time-consuming**
- Some taxonomies needs to **be updated continuously**
- **Low coverage & inconsistency** of taxonomy will hurt the performance of downstream applications

Solution

- Accords to the rules while reducing human effort: **Expand a pre-defined seed taxonomy** rather than construct from scratch
- Updating taxonomy in a system: Concept mining → taxonomy expansion → item tagging in a row

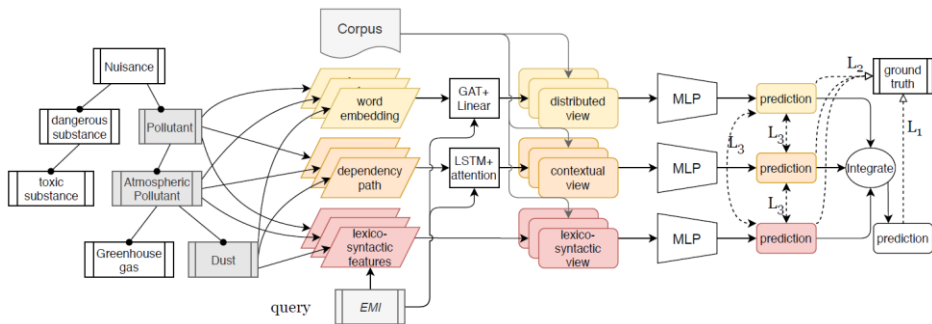
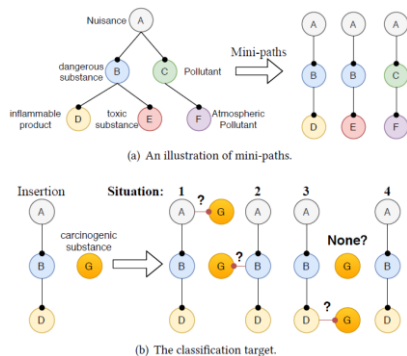
TaxoExpan: Self-supervised Taxonomy Expansion with Position-Enhanced Graph Neural Network (WWW '20)



Modeling

- A mere node attaching algorithm for general purpose taxonomy
- Taxonomy modeled as a Bayesian network, do not have to be a tree
- Features of anchor node captured by Positional GAT on "egonet" containing its direct parents and children
- Score by a matching module, whose input is concatenation of query representation (embedding) & anchor representation (GAT results)
- Dataset constructed with self-supervision

STEAM: Self-supervised Taxonomy Expansion with Mini-paths (SIGKDD '20)

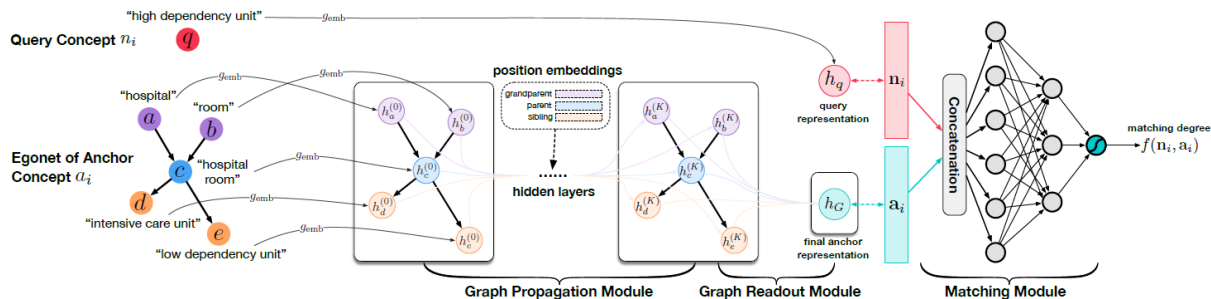
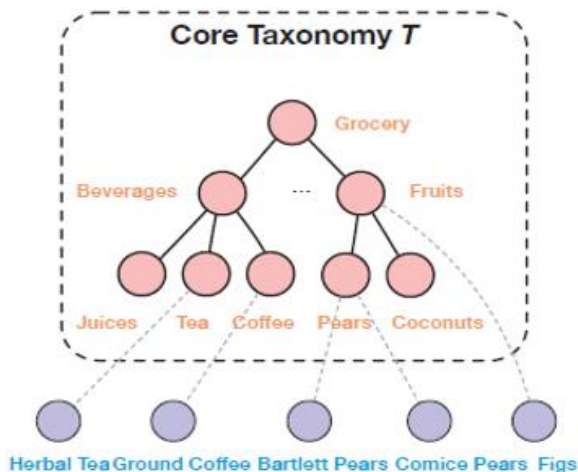


Modeling

- A mere node attaching algorithm for general purpose taxonomy
- Taxonomy modeled as a **tree** (not a Bayesian network)
- Extract **distributed feature** by **Positional GAT** with **BERT embedding** as input
- Extract **contextual feature** by **weighted-summing** the **LSTM** outputs of the **dependency trees** of query-anchor co-occurrence in corpus
- Extract **lexical-syntactic feature** by measuring a series of **lexical relationship features**
- Loss is the sum of **aggregated loss** containing 1) classifier loss, 2) single classifier loss and 3) cross-classifier consistency loss
- Dataset constructed with self-supervision

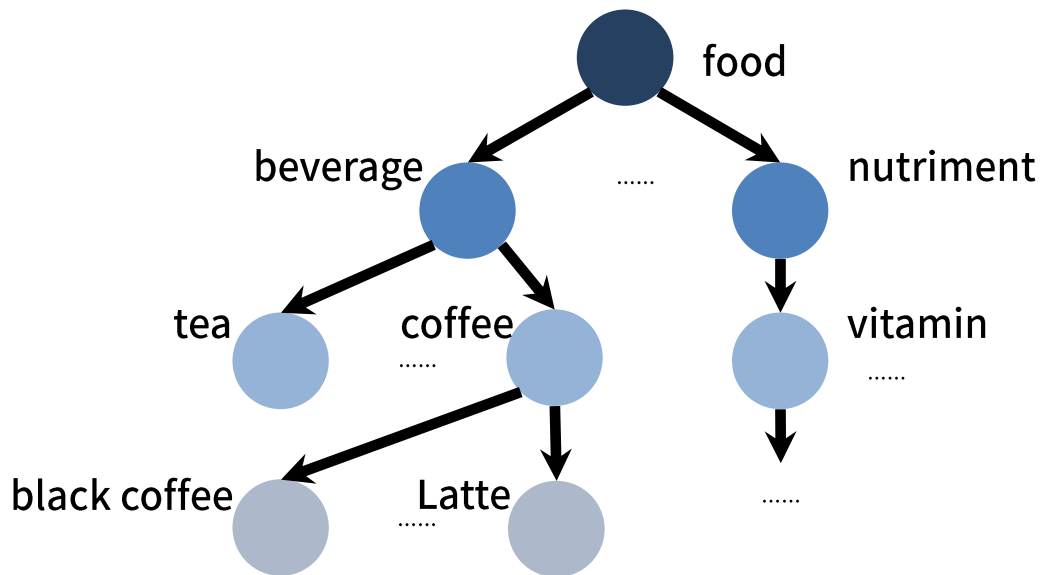
Overall Problems in Previous Solutions

- Coarsely designed lexical and syntactical features: only extract superficial features in node names
- Deficient usage of other text data source: no definition texts are properly used in previous methods
- Insufficient mining of tree structure: learn hypernym detection of a single node, not construction of a hierarchy
- Ignorance of cross-level relationships: the nodes on the right path in taxonomy should share similarity
- High time complexity: need to compare all possible nodes to achieve the result ranking



How to Add a New Term to a Taxonomy?

Seed Taxonomy

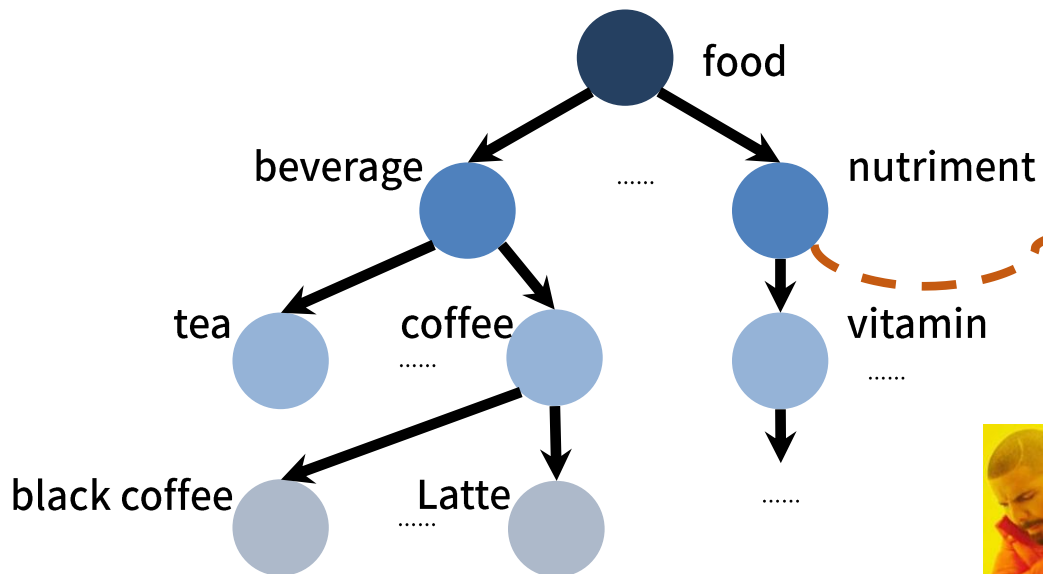


New Query Term

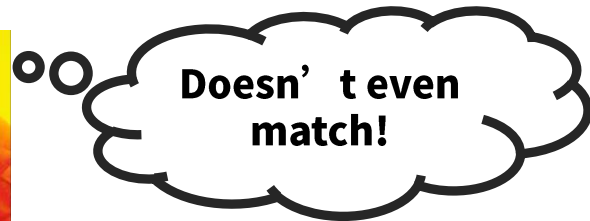
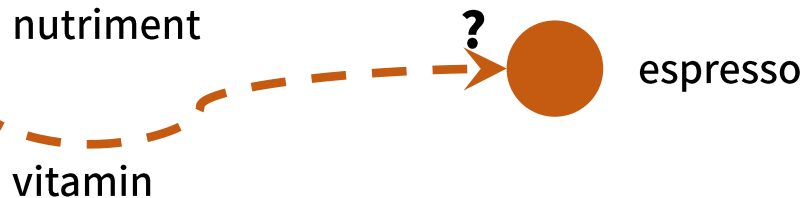


How to Add a New Term to a Taxonomy?

Seed Taxonomy



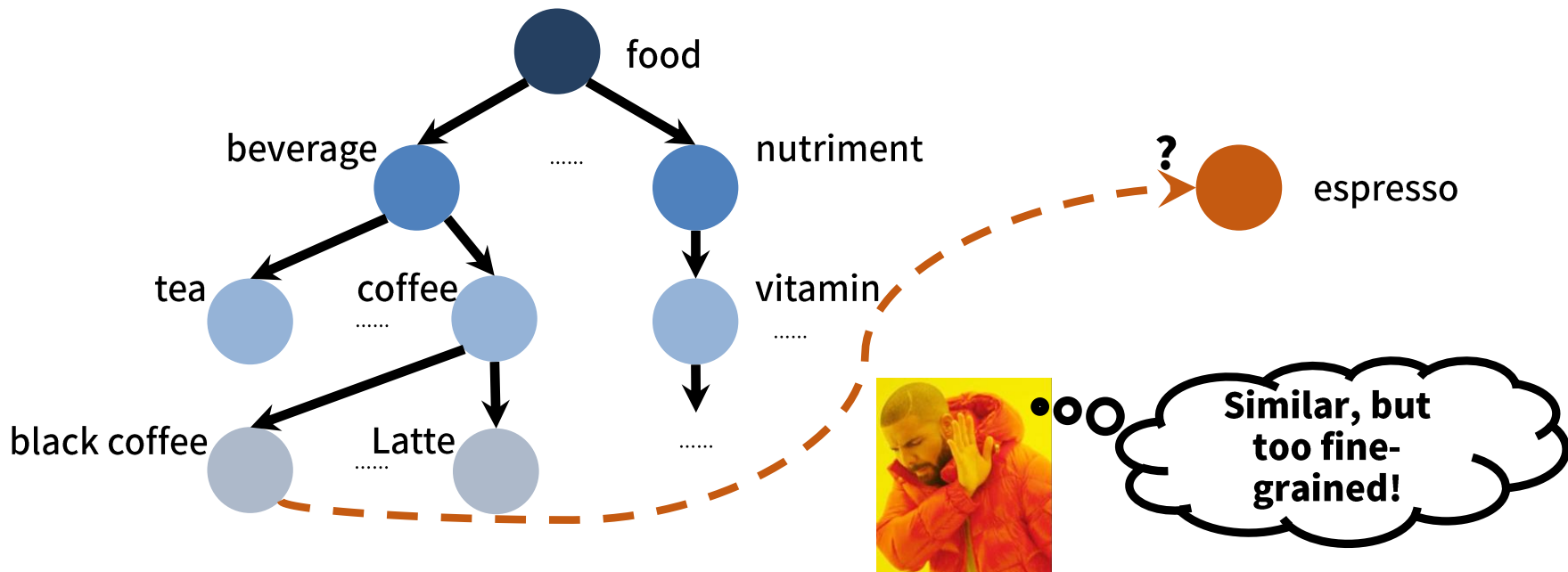
New Query Term



How to Add a New Term to a Taxonomy?

Seed Taxonomy

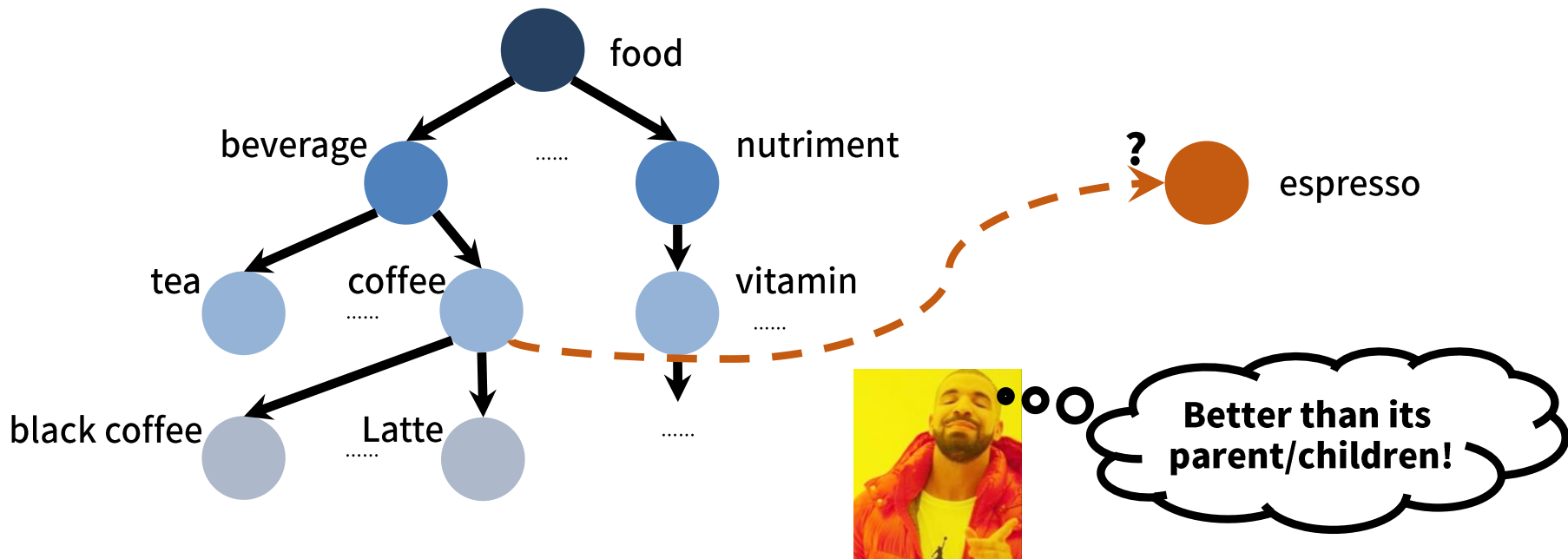
New Query Term



How to Add a New Term to a Taxonomy?

Seed Taxonomy

New Query Term



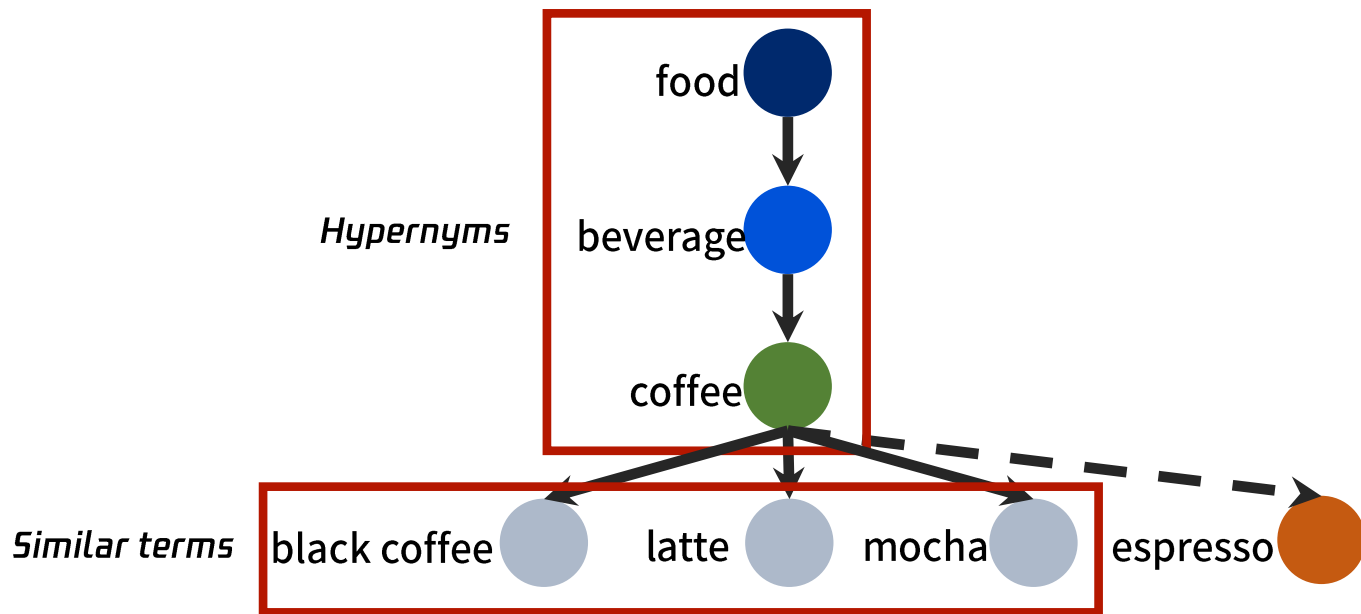
Motivation of Our Model



Term's *surface name* is *insufficient* for hypernymy detection

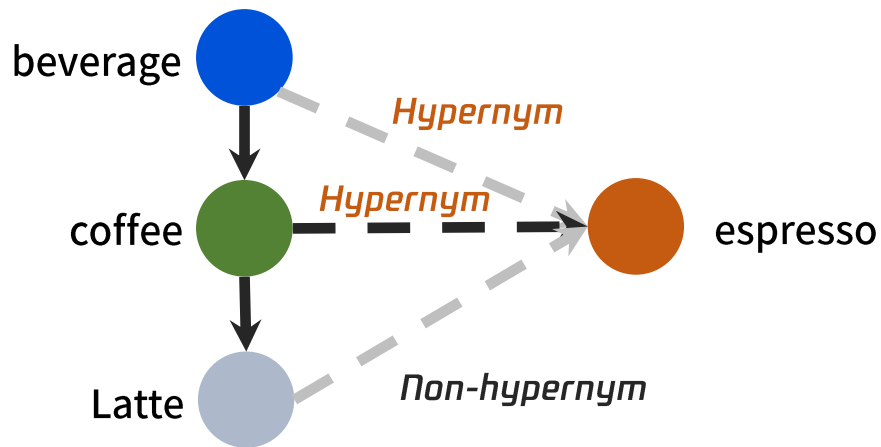
Contextual pattern in large corpora includes *noice*

— Motivation of Our Model



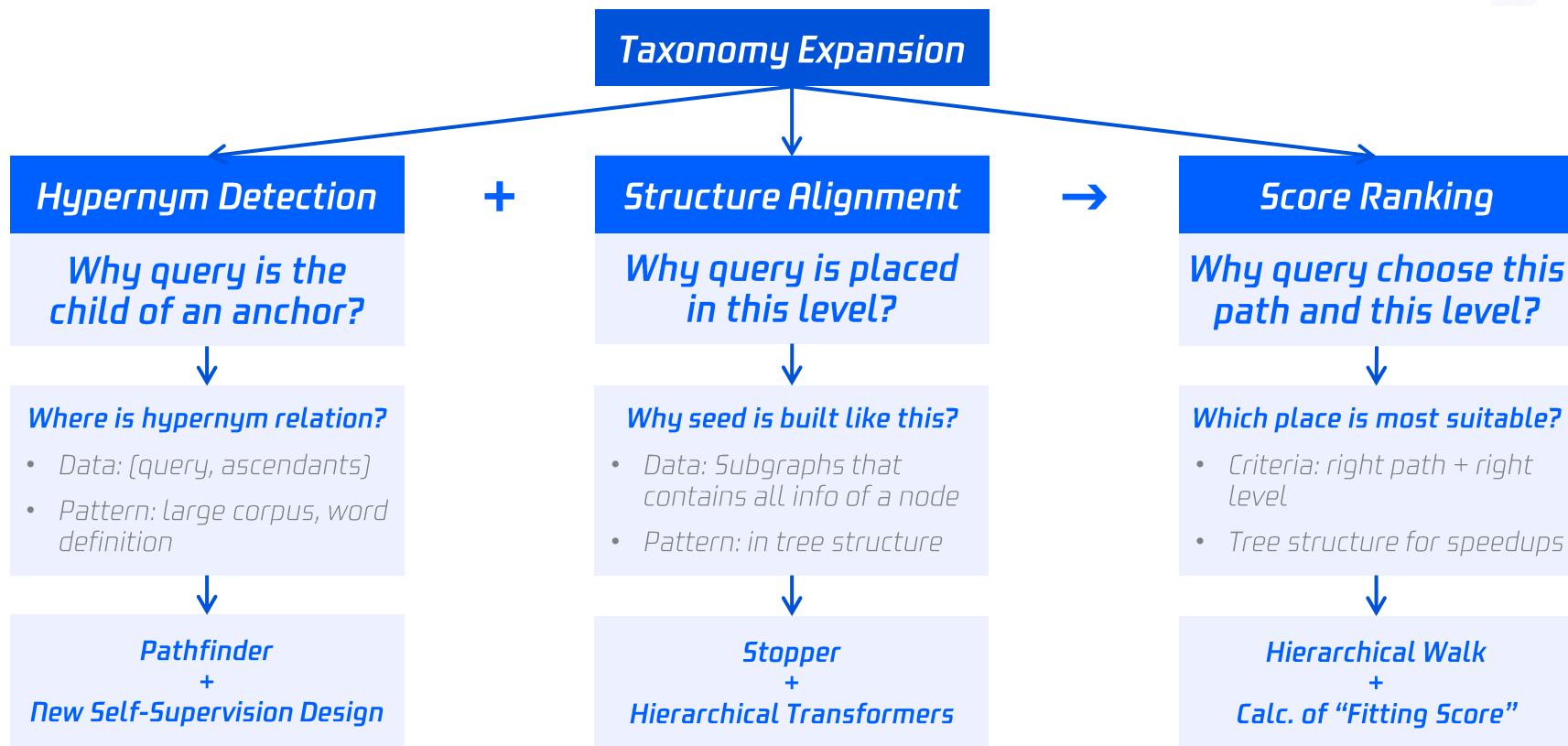
Hypernymy relations from parent-child relations
+ *similarity relations* from sibling relations = *tree coherence*

— Motivation of Our Model



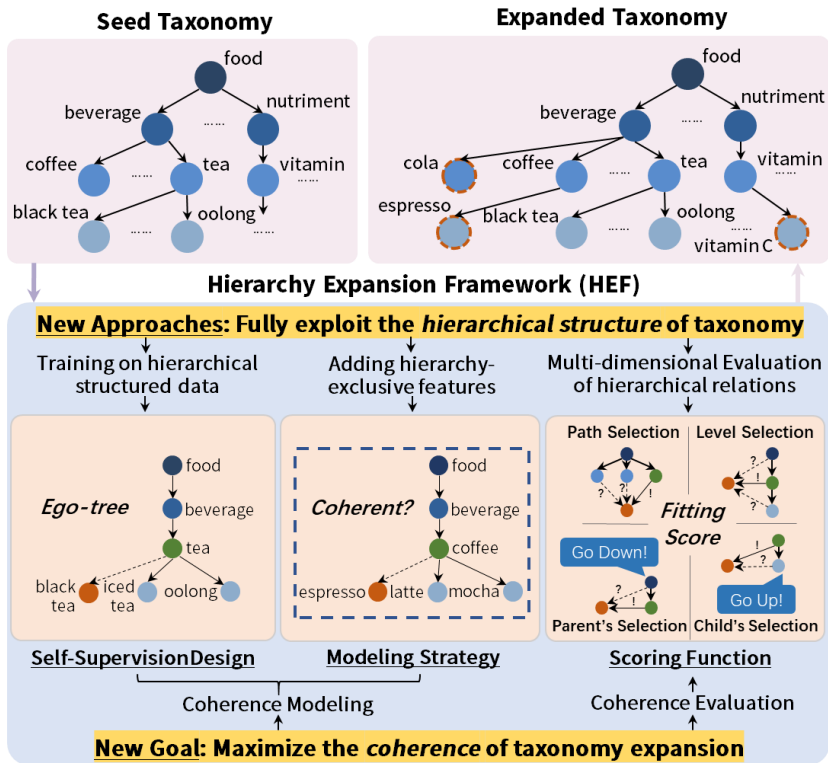
*From “is-A” relation to “The most accurate is-A” relation
needs **comparison with parent/children***

Deconstruction of the Taxonomy Expansion Task

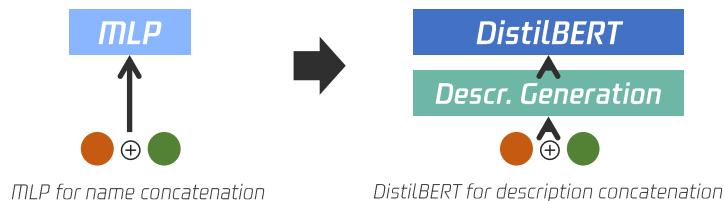


Motivation and Novelties

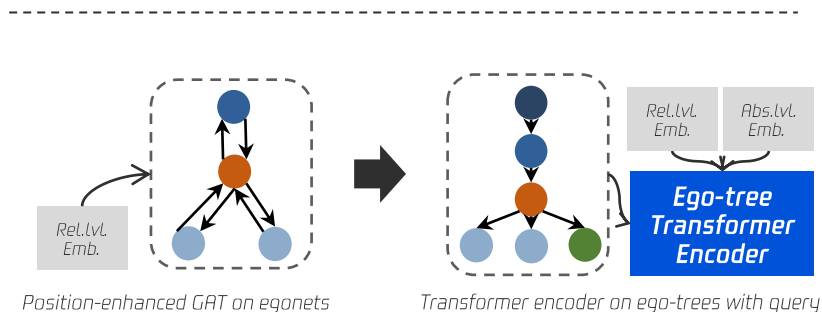
Major: New approaches and new goal



Others: New Implementations

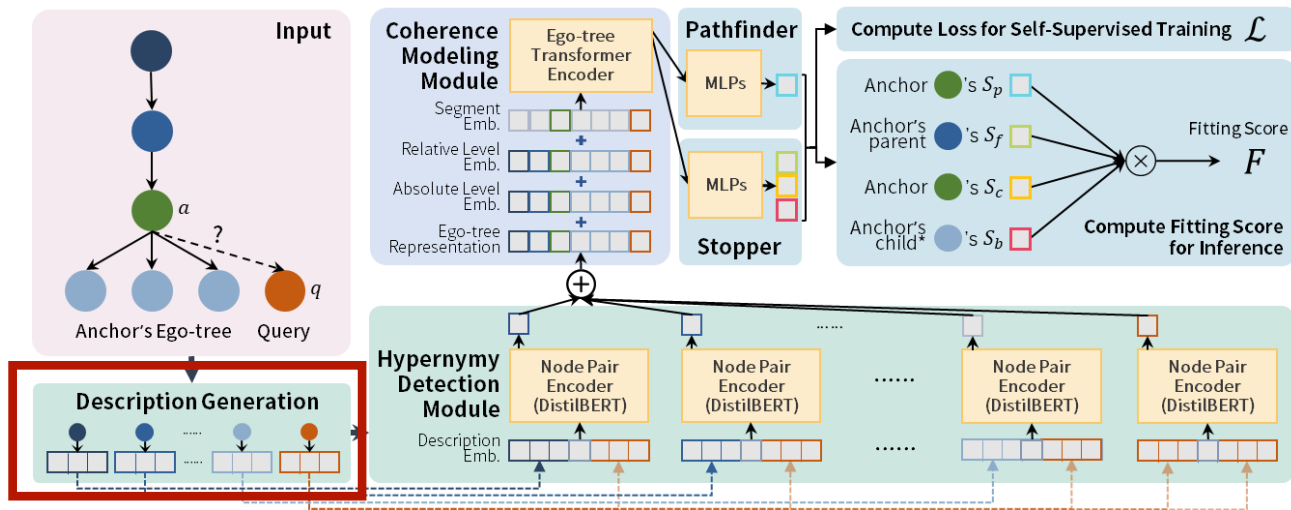


Node pair relation detection



Structure modeling

Model Design – Description Generation



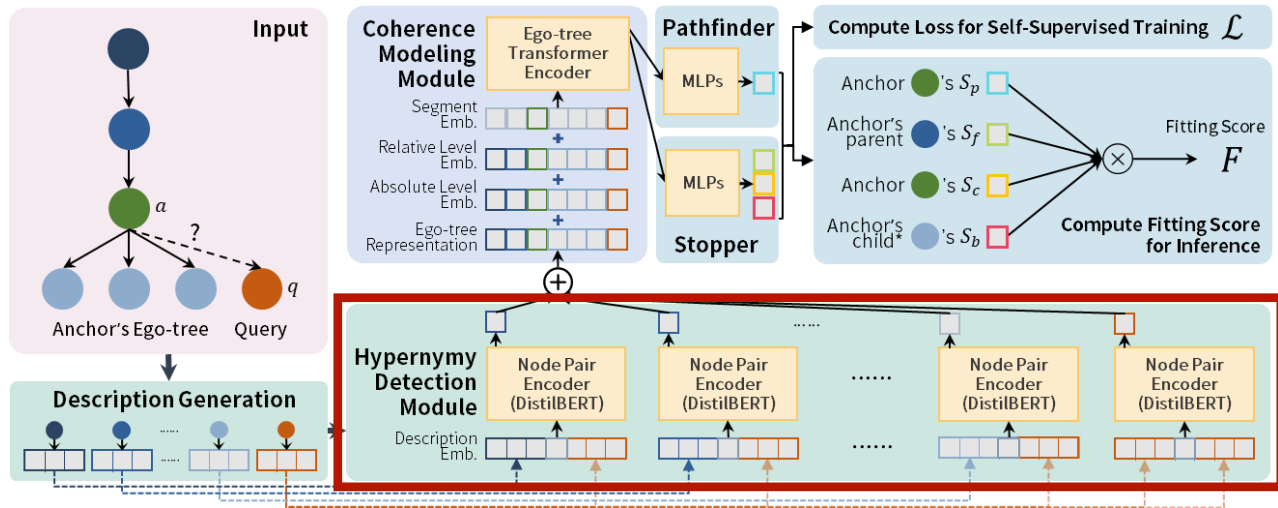
“Adaptation to Climate Change”



algorithm based on DP and WordNet

“adaptation to climate change is the process of adapting to something (such as environmental conditions) to a change in the world's climate”

Model Design – Hypernymy Detection Module



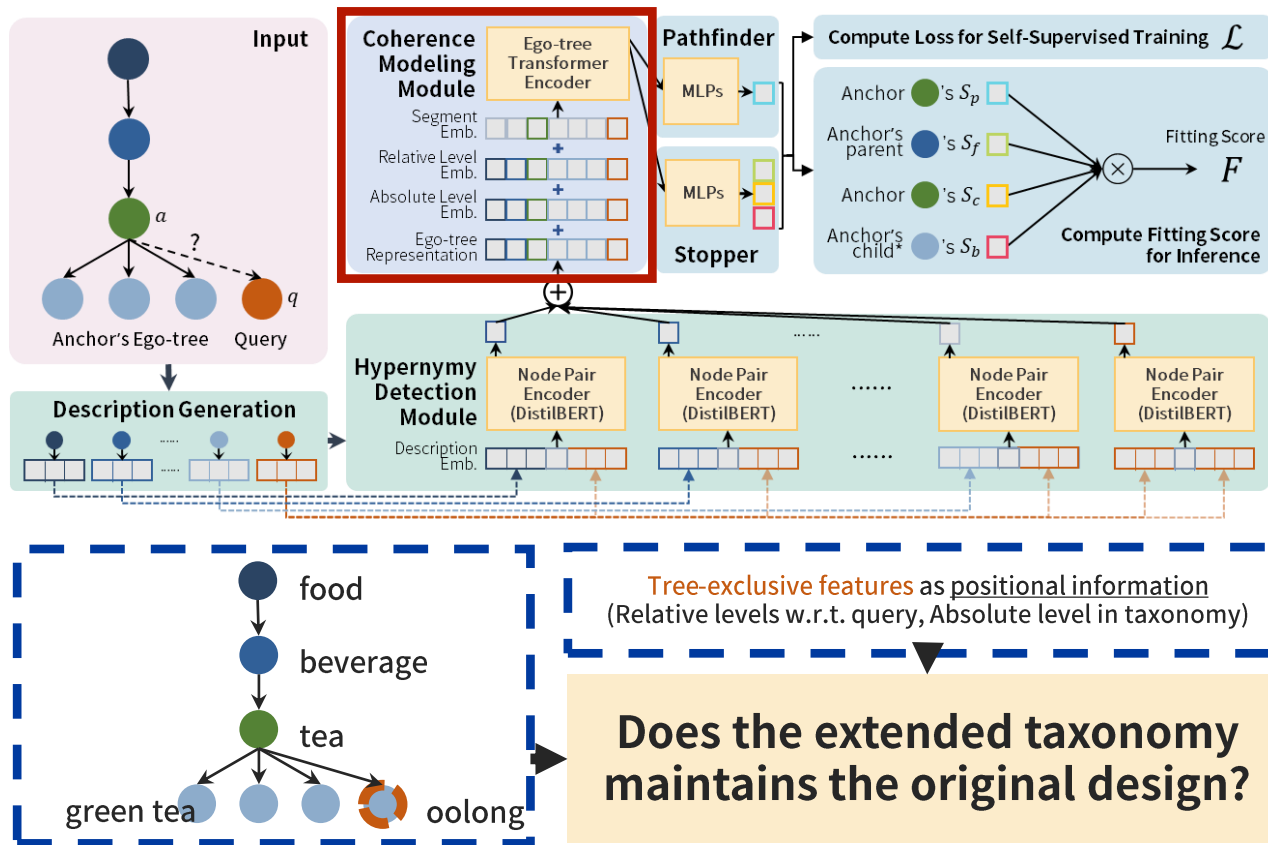
“oolong is Chinese tea leaves that have been partially femented...”



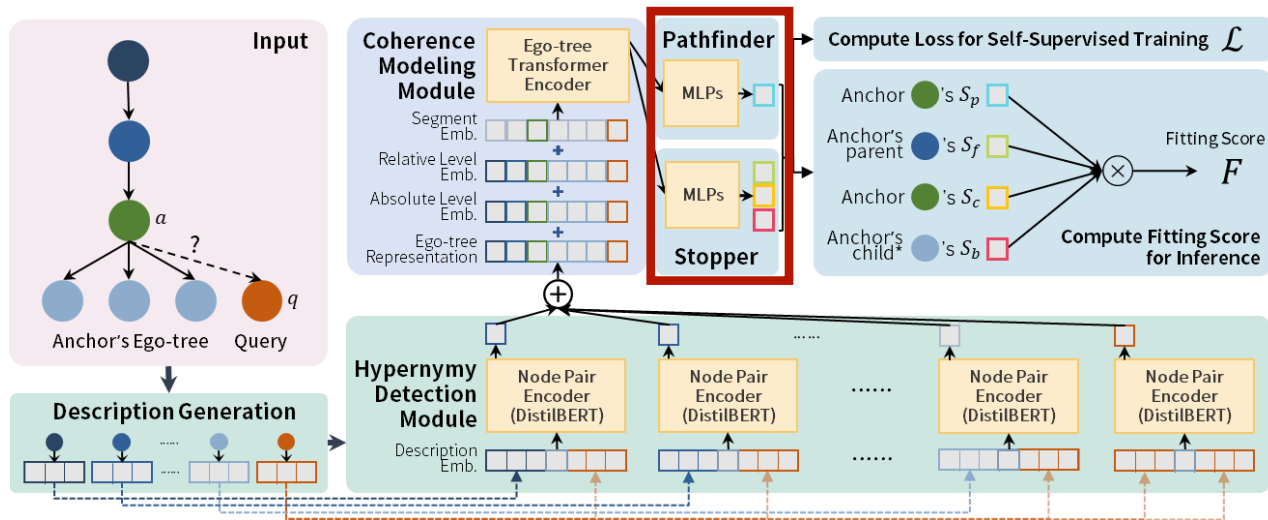
DistilBERT for hypernymy detection

“tea is a tropical evergreen shrub or small tree extensively cultivated...”

Model Design – Coherence Modeling Module



Model Design – Pathfinder & Stopper



Tree structure enables **dual evaluation**:

Pathfinder (Path evaluation)

S_p for path evaluation (hypernymy relation?)

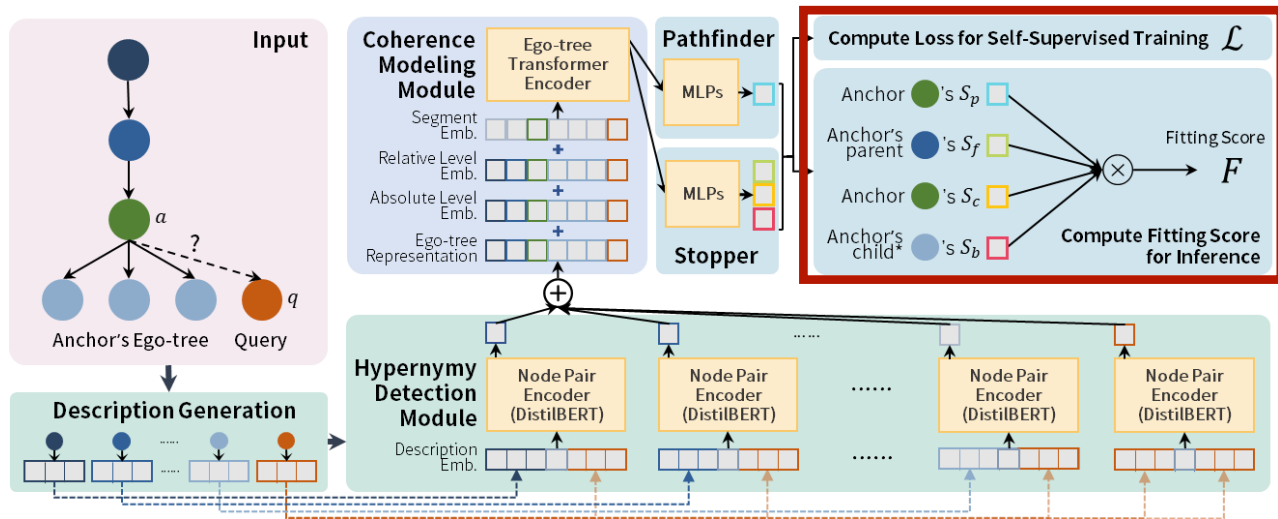
Stopper (Level evaluation + comparison)

S_f for suggestion as parent (child may be better)

S_c for level evaluation (suitable level?)

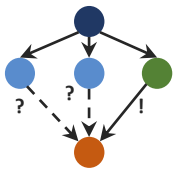
S_b for suggestion as children (parent may be better)

Model Design – Pathfinder & Stopper

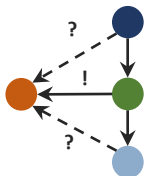


Fitting score integrating **Path + Level, Parent + Children** Selection

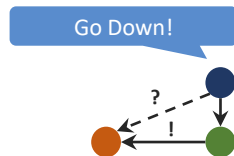
Path Selection



Level Selection



Parent's Selection



Child's Selection



Results of Main Experiments and Ablation Studies

	SemEval-16 Task 13 - Environment			SemEval-16 Task 13 - Science			SemEval-16 Task 13 – Food		
Methods	Acc	MRR	Mean Wu-Palmer	Acc	MRR	Mean Wu-Palmer	Acc	MRR	Mean Wu-Palmer
BERT+MLP	11.1	21.5	47.9	11.5	15.7	43.6	10.5	14.9	47.0
TAXI (ACL '16)	16.7	-	44.7	13.0	-	32.9	18.2	-	39.2
HypeNet (ACL '16)	16.7	23.7	55.8	15.4	22.6	50.7	20.5	27.3	63.2
TaxoExpan (WWW '20)	11.1	32.3	54.8	27.8	44.8	57.6	27.6	40.5	54.2
STEAM (SIGKDD '20)	36.1	46.9	69.6	36.5	48.3	68.2	34.2	43.4	67.0
HEF	55.3	65.3	71.4	53.6	62.7	75.6	47.9	55.5	73.5

SemEval-16 Task 13 Environment	Acc	MRR	Mean Wu-Palmer
HEF	55.3	65.3	71.4
- WordNet Descr.	41.5	55.3	62.6
- Ego-tree + Egonet	45.3	60.6	69.9
- Relative Level Emb.	49.1	59.2	60.9
- Absolute Level Emb.	49.1	60.6	68.4
Stopper Only	52.8	62.5	68.7
Pathfinder + Current Only	50.9	62.1	66.8
Current Only	41.5	54.7	58.6

	Environment	Science	Food
Nodes	261	429	1486
Edges	260	428	1485
Depth	6	8	8

$$Acc = Hit@1 = \frac{1}{n} \sum_{i=1}^n \mathbb{I}(y_i = \hat{y}_i)$$

$$MRR \text{ (Mean Reciprocal Rank)} = \frac{1}{n} \sum_{i=1}^n \frac{1}{\{rank(y_i)\}}$$

$$\text{Mean Wu-Palmer Similarity} = \frac{1}{n} \sum_{i=1}^n \frac{2 \times \text{depth}(LCA(y_i, \hat{y}_i))}{\text{depth}(y_i) + \text{depth}(\hat{y}_i)}$$

Case Studies and Future Works

Query (q)	Ground Truth (\hat{p})	Scores	Prediction (p)	Scores
q : paddy is rice in the husk either gathered or still in the field	\hat{p} : rice is grains used as food either unpolished or more often polished	$S_p = 0.9997$ $S_c = 0.4599$	p : rice is grains used as food either unpolished or more often polished	$S_p = 0.9997$ $S_c = 0.4599$
	parent(\hat{p}): starches is a commercial preparation of starch that is used to stiffen textile fabrics in laundering	$S_f = 0.9755$	parent(p): starches is a commercial preparation of starch that is used to stiffen textile fabrics in laundering	$S_f = 0.9755$
$F(\hat{p}, q) = 0.4483$ \hat{p} 's Ranking: 1	c_p^* : white rice is having husk or outer brown layers removed	$S_b = 0.9995$	c_p^* : white rice is having husk or outer brown layers removed	$S_b = 0.9995$
q : fish meal is ground dried fish used as fertilizer and as feed for domestic livestock	\hat{p} : feed is food for domestic livestock	$S_p = 0.9993$ $S_c = 0.3169$	p : feed is food for domestic livestock	$S_p = 0.9993$ $S_c = 0.3169$
	parent(\hat{p}): food is any substance that can be metabolized by an animal to give energy and build tissue	$S_f = 0.9984$	parent(p): food is any substance that can be metabolized by an animal to give energy and build tissue	$S_f = 0.9984$
$F(\hat{p}, q) = 0.3158$ \hat{p} 's Ranking: 1	c_p^* : mash is mixture of ground animal feeds	$S_b = 0.9988$	c_p^* : mash is mixture of ground animal feeds	$S_b = 0.9988$
q : bourguignon is reduced red wine with onions and parsley and thyme and butter	\hat{p} : sauce is flavorful relish or dressing or topping served as an accompaniment...	$S_p = 0.0002$ $S_c = 0.0001$	p : wine is a red as dark as red wine	$S_p = 0.9997$ $S_c = 0.1399$
	parent(\hat{p}): condiment is a preparation (a sauce or relish or spice) to enhance flavor or enjoyment	$S_f = 0.0004$	parent(p): alcohol is any of a series of volatile hydroxyl compounds that are made from hydrocarbons by distillation	$S_f = 0.9812$
$F(\hat{p}, q) = 1e - 11$ \hat{p} 's Ranking: 328	c_p^* : bercy is butter creamed with white wine and shallots and parsley	$S_b = 0.9997$	c_p^* : red wine is wine having a red color derived from skins of dark-colored grapes	$S_b = 0.8784$
q : hot fudge sauce is hot thick chocolate sauce served hot	\hat{p} : chocolate sauce is sauce made with unsweetened chocolate or cocoa...	$S_p = 0.9471$ $S_c = 9e - 5$	p : sauce is flavorful relish or dressing or topping served as an accompaniment...	$S_p = 0.9995$ $S_c = 0.0172$
	parent(\hat{p}): sauce is flavorful relish or dressing or topping served as an accompaniment...	$S_f = 0.9617$	parent(p): condiment is a preparation (a sauce or relish or spice) to enhance flavor or enjoyment	$S_f = 0.9888$
$F(\hat{p}, q) = 6e - 6$ \hat{p} 's Ranking: 20	c_p^* : None	$S_b = 0.0700$	c_p^* : lyonnaise sauce is brown sauce with sauteed chopped onions and parsley...	$S_b = 0.9995$

3rd: Bourguignon is also a kind of sauce for cooking meat. The **description** is incorrect.

4th: Leaf nodes' backward scores are assigned to be a constant. The **backward score assignment** is incorrect.

What's Next?

No WordNet descriptions for Chinese?: Discover approaches for generating useful descriptions suggesting hypernymy relations for Chinese concepts

New ways for integrating larger corpus: Find ways to use larger corpus or query logs to boost the performance of hypernymy detection / coherence modeling

...and to fix the bad cases:

Better description generation algorithm: Develop a DP+ML algorithm to generate domain-related descriptions for terms

Backward scores for leaf node's inexistent child: New approaches for obtaining the backward scores

Some useful links

Our paper: <https://arxiv.org/abs/2101.11268>

An awesome list for taxonomy-related research:

<https://github.com/mickeystroller/awesome-taxonomy>

Thanks