

Agenda

08:30 PST	10 mins	Introduction to commonsense knowledge (Filip)
08:40 PST	25 min	Part I - Axiomatization of commonsense knowledge (Mayank)
09:05 PST	40 min	Part II - Consolidating commonsense knowledge (Filip)
09:45 PST	15 min	Break
10:00 PST	45 min	Part III - Extracting and contextualizing commonsense knowledge (Simon)
10:45 PST	45 min	Part IV - Language models, QA, and evaluation challenges (Antoine)
11:30 PST	15 min	Way forward: KGs+LMs+axioms? (Filip)

Consolidating Commonsense Knowledge

Filip Ilievski

Overview of CSK sources

Category	Source	Relations	Example 1	Example 2
Commonsense KGs	ConceptNet*	34	<i>food - capable of - go rotten</i>	<i>eating - is used for - nourishment</i>
	ATOMIC	9	<i>Person X bakes bread - xEffect - eat food</i>	<i>PersonX is eating dinner - xEffect - satisfies hunger</i>
	GLUCOSE	10	<i>Someone_A makes Something_A (that is food)</i>	<i>Causes/Enables Someone_A eats Something_A</i>
	WebChild	4 (groups)	<i>restaurant food - quality#n#1 - expensive</i>	<i>eating - type of - consumption</i>
	Quasimodo	78,636	<i>pressure cooker - cook faster - food</i>	<i>herbivore - eat - plants</i>
	SenticNet	4	<i>cold_food - polarity - negative</i>	<i>eating breakfast - polarity - positive</i>
	HasPartKB	1	<i>dairy food - has part - vitamin</i>	<i>n/a</i>
Common KGs	Wikidata	6.7k	<i>food - has quality - mouthfeel</i>	<i>eating - subclass of - ingestion</i>
	YAGO4	116	<i>banana chip - rdf:type - food</i>	<i>eating - rdfs:label - feeding</i>
	DOLCE*	1	<i>n/a</i>	<i>n/a</i>
	SUMO*	1,614	<i>food - hyponym - food_product</i>	<i>process - subsumes - eating</i>
Lexical resources	WordNet	10	<i>food - hyponym - comfort food</i>	<i>eating - part-meronym - chewing</i>
	Roget	2	<i>dish - synonym - food</i>	<i>eating - synonym - feeding</i>
	FrameNet	8 (f2f)	<i>Cooking_creation - has frame element - Produced_food</i>	<i>eating - evoke - Ingestion</i>
	MetaNet	14 (f2f)	<i>Food - has role - food_consumer</i>	<i>consuming_resources - is - eating</i>
	VerbNet	36 (roles)	<i>feed.v.01 - Arg1-PPT - food</i>	<i>eating - hasPatient - comestible</i>
Visual sources	Visual Genome	42,374	<i>food - on - plate</i>	<i>boy - is eating - treat</i>
	Flickr30k	1	<i>a food buffet - corefers with - a food counter</i>	<i>a eating place - corefers with - their kitchen</i>
Corpora & LMs	GenericsKB	n/a	<i>Aardvarks search for food.</i>	<i>Animals receive nitrogen by eating plants.</i>
	GPT-2	n/a	<i>Food causes a person to be hungry and a person to eat.</i>	<i>Eating at home will not lead to weight gain.</i>

Consolidation Hypothesis

Integrating multiple knowledge sources in CSKG is beneficial for downstream reasoning tasks.

On stage, a woman takes a seat at the piano. She

1. sits on a bench as her sister plays with the doll.
2. smiles with someone as the music plays.
3. is in the crowd, watching the dancers.
4. nervously sets her fingers on the keys.

piano is used for...

- en performing music →
- en music →
- en accompanying an orchestra →

Things located at piano

- en keys →
- en black keys →
- en hammers →
- en a keyboard →

ConceptNet: pianos have keys, are used to perform music

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Visual Genome: person can play a piano while sitting, his hands are on the keyboard

man plays piano
keys ON piano
woman watches man
pillow ON couch
light ON wall
window IN room
person playing piano
guy ON bench
hands ON keyboard

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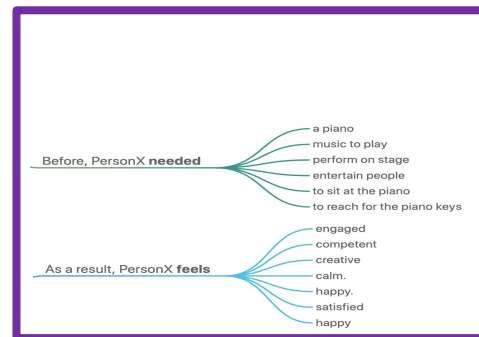
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ATOMIC: to play piano, a person needs to sit at it, on stage and reach for the keys; feelings

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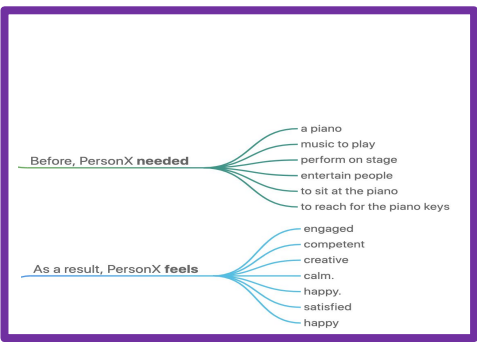
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FrameNet: performer entertains audience

Audience [Aud]	The Audience experiences the Performance .
Medium [Medium]	Medium is the physical entity or channel used by the Performer to transmit the Performance to the Audience .
Performance [Perance]	The Performers generates the Performance which the Audience perceives.
Performer [Perfer]	The Performer provides an experience for the Audience .

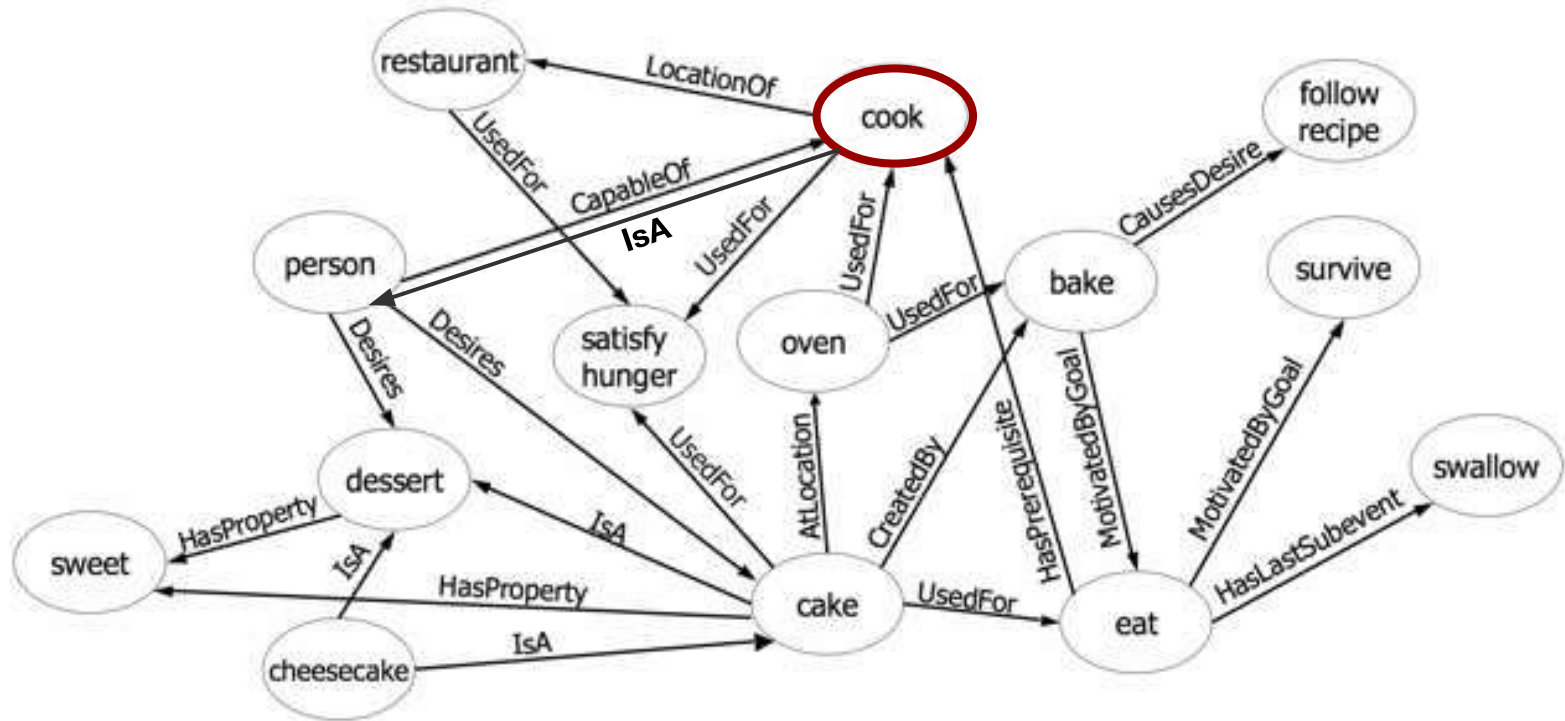
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Challenge: Knowledge granularity

	size	examples
Concept Net	36 relations, 8M nodes, 21M edges	/c/en/piano /c/en/piano/n /c/en/piano/n/wn /r/relatedTo
Web Child	4 relation groups, 2M nodes, 18M edges	hasTaste fasterThan
ATOMIC	9 relations, 300k nodes, 877k edges	wanted-to impressed
Wikidata	1.2k relations, 75M objects, 900M edges	wd:Q1234 wdt:P31
CEO	121 properties, 223 events	ceo:Damaging hasPostSituation
WordNet	10 relations, 155k words, 176k synsets	dog.n.01 hypernymy
Roget	2 relations, 72k words, 1.4M edges	truncate antonym
VerbNet	273 top classes 23 roles, 5.3k senses	perform-v performance-26.7-1
FrameNet	1.9k edges, 1.2k frames, 12k roles, 13k lexical units	Activity Change_of_leadership New_leader
Visual Genome	42k relations, 3.8M nodes, 2.3M edges, 2.8M attributes	fire hydrant white dog

Challenge: Imprecise descriptions



Challenge: Sparse overlap and mappings

	Other source	Both Wikidata-CS	only Other source only
ConceptNet	2,386	97,473 (97.6%)	3,320,935 (99.9%)
Roget	299	99,560 (99.7%)	1,403,162 (99.9%)
WordNet	1,613	98,246 (98.4%)	419,103 (99.6%)

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	mappings
Concept Net	WordNet, DBpedia, OpenCyc, Wiktionary
Web Child	WordNet
ATOMIC	ConceptNet, Cyc
Wikidata	various
CEO	FrameNet, SUMO
WordNet	
Roget	
VerbNet	FrameNet, WordNet
FrameNet	
Visual Genome	WordNet

Challenge: Modeling of relations

ConceptNet

Web Child

/r/HasProperty



-
ability#n#1
age#n#1
appearance#n#1
beauty#n#1
color#n#1
disposition#n#4
emotion#n#1
feeling#n#1
length#n#1
manner#n#1
motion#n#4
personality#n#1
physical_property#n#1

quality#n#1
sensitivity#n#2
shape#n#2
size#n#1
sound#n#1
state#n#2
strength#n#1
structure#n#2
sustainability#n#1
tactile_property#n#1
taste_property#n#1
temperature#n#1
trait#n#1
weight#n#1

Challenge: Acquisition methods and quality

	creation
Concept Net	crowd- sourcing
Web Child	curated automatic extraction
ATOMIC	crowd- sourcing
Wikidata	crowd- sourcing
CEO	manual
WordNet	manual
Roget	manual
VerbNet	manual
FrameNet	manual
Visual Genome	crowd- sourcing

Consolidation of nodes

P1. Embrace heterogeneity of nodes

objects, classes, words, actions, frames, states

P2. Leverage external links

many sources map to WordNet

P3. Generate high-quality probabilistic links

many facts not explicitly stated

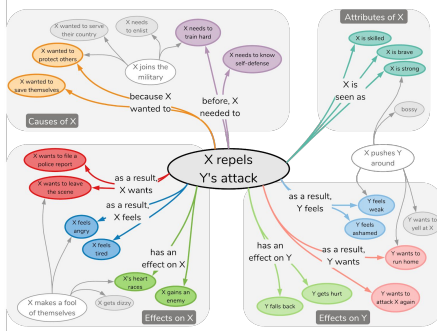
Overview of node mappings

251,517 mw:SameAs

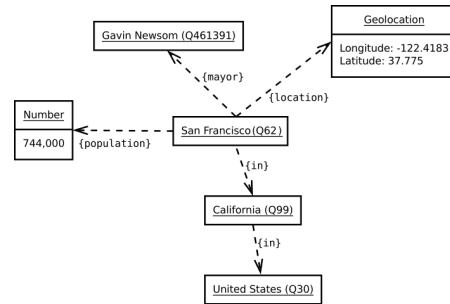
45,659 mw:HasInstance

Mapping from	Mapping to	Relation	Resource used
WordNet 3.0 senses	WordNet 3.1 senses	mw:SameAs	Interlingual Index (ILI)
lexical nodes in ConceptNet	lexical nodes in ATOMIC and ROGET	mw:SameAs	/
ConceptNet nodes	FrameNet LUs	mw:SameAs	Predicate matrix
ConceptNet concepts	FrameNet FEe	mw:HasInstance	rule-based system
Wikidata Qnodes	WordNet senses	mw:SameAs	XLNet-based description similarity

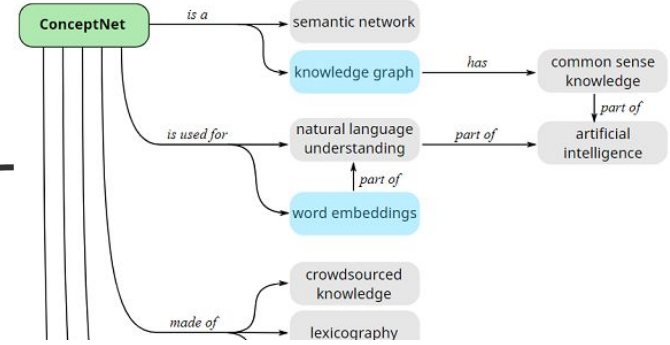
Consolidated Knowledge Graphs



ATOMIC (Sap et al. 2019)



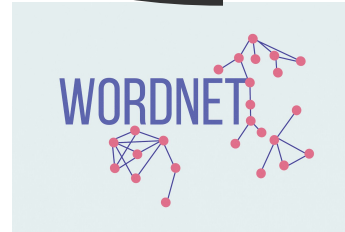
Wikidata (Vrandecic and Krotzsch 2014)



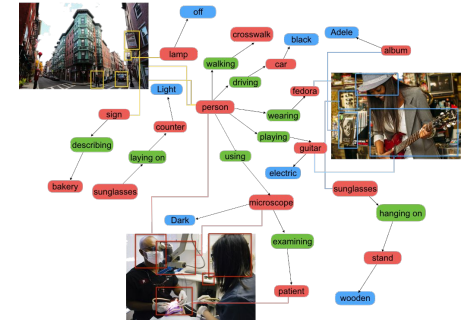
ConceptNet (Speer, Chin and Havasi 2017)



CSKG (Ilievski et al. 2020)

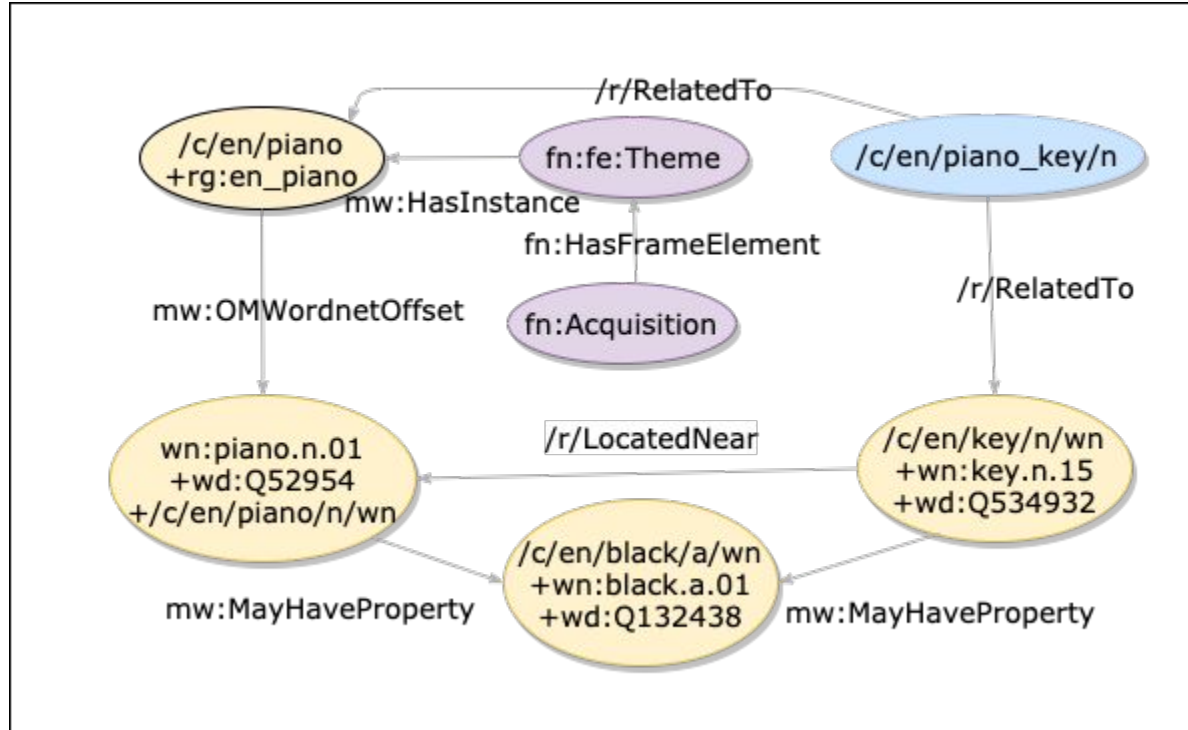


WordNet (Miller 1995)

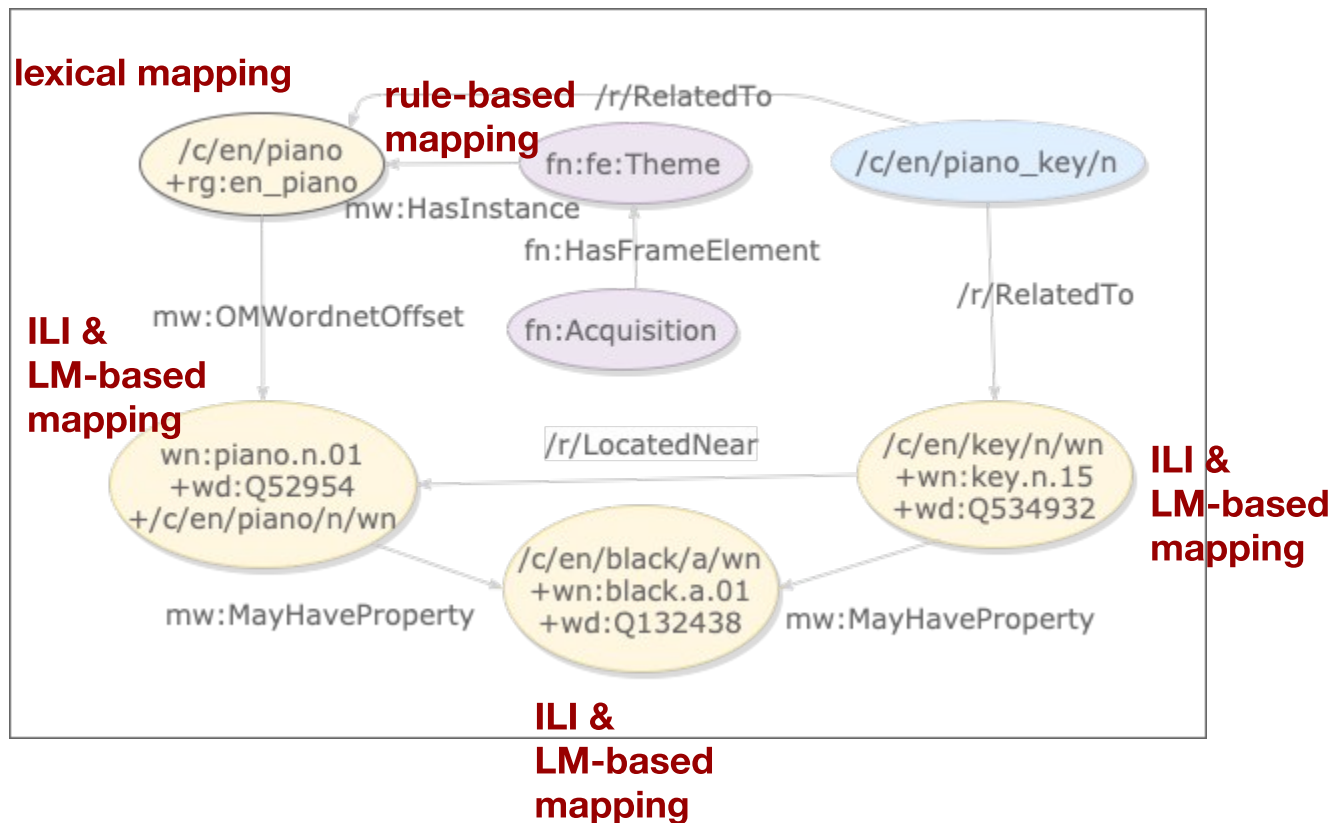


Visual Genome (Krishna et al. 2017)

CSKG snippet



The effect of node resolution



Consolidation of relations

P1. Reuse edge types across resources

-> 58 relations

/r/LocatedNear from ConceptNet applicable for attributes in Visual Genome

Integration statistics

	AT	CN	FN	RG	WN	WD	VG	CSKG (concat)	CSKG
#nodes	304,909	1787373	15,652	71,804	91,294	71,243	11,264	2,414,813	2,160,968
#edges	732,723	3,423,004	54,109	1,403,955	111,276	101,771	2,587,623	6,349,731	6,001,531
#relations	9	34	23	2	3	15	3	59	58
mean degree	4.81	3.83	6.91	39.1	2.44	2.44	459.45	5.26	5.55

Top PageRank nodes

1. /c/en/**chemical_compound**/n
2. /c/en/**change**/n/wn/artifact
3. /c/en/**natural_science**/n/wn/cognition
4. /c/en/**chromatic**/a/wn
5. /c/en/**organic_compound**

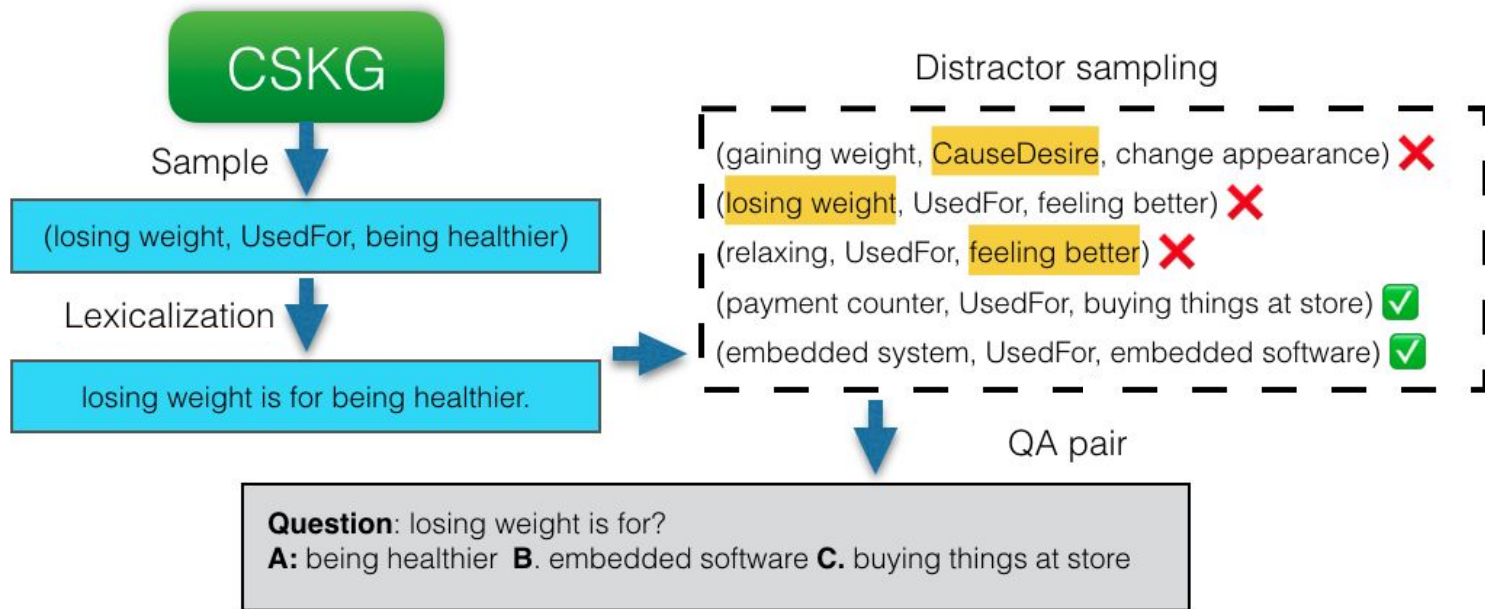
Zero-shot QA with CSKG

Method

(Ma et al., 2021)

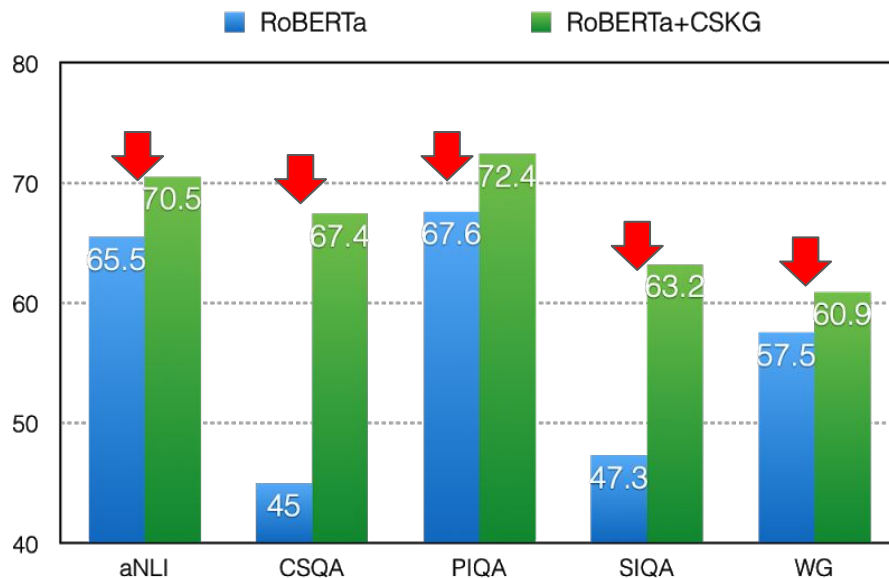
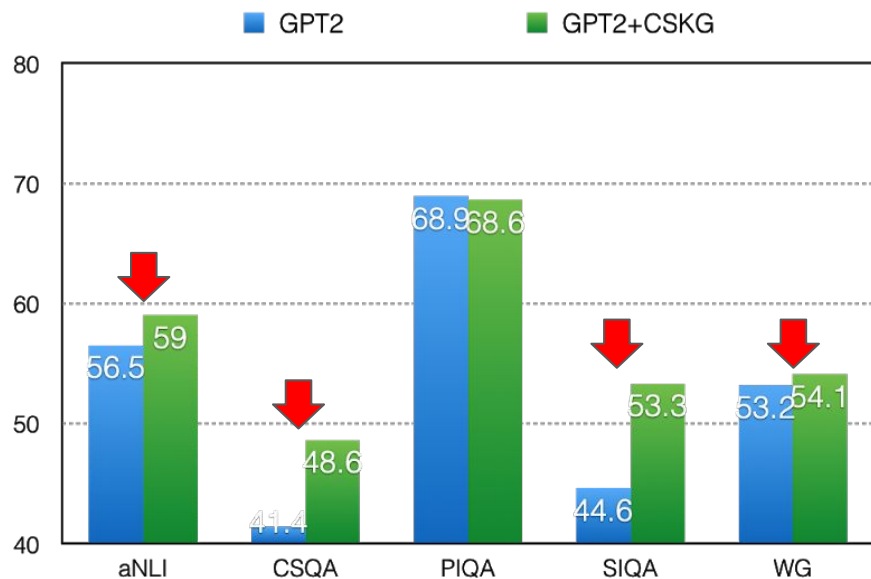
Pretrain LMs with artificial QA sets generated from CSKG

Answer commonsense questions on unseen datasets



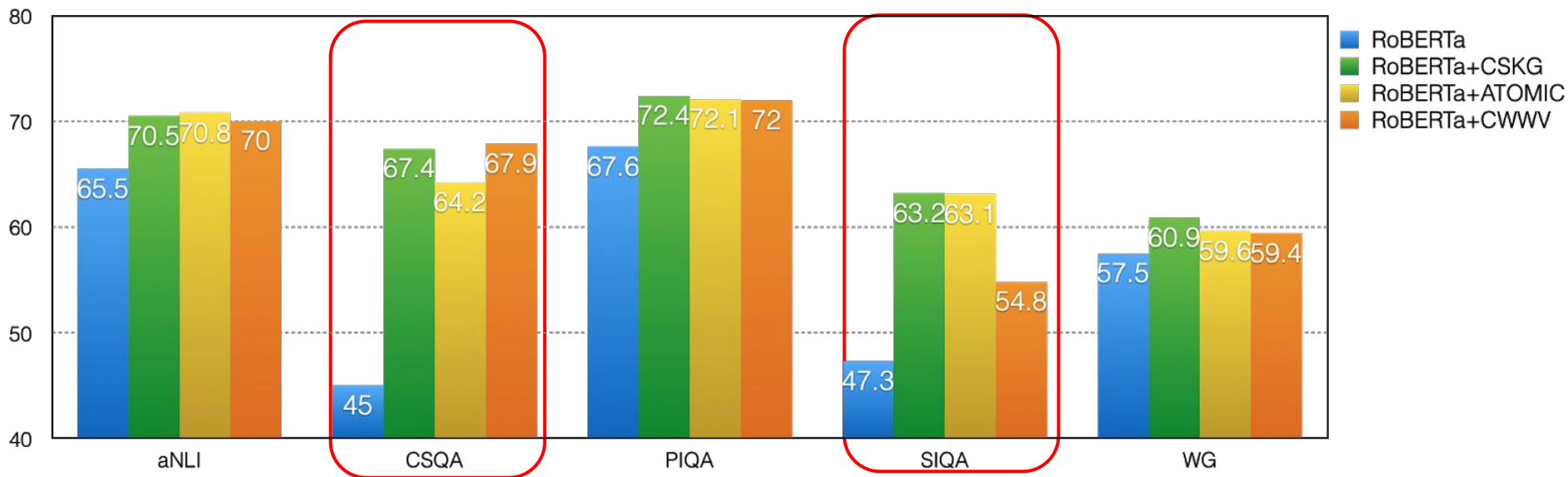
Main Results

- Pretraining on artificial QA sets helps accuracy



Main Results

- The impact of knowledge depends on KG-task alignment
- Adding knowledge improves accuracy



Consolidation of relations

P1. Reuse edge types across resources

/r/LocatedNear from ConceptNet applicable for attributes in Visual Genome

P2. Group relations into high-level **dimensions**

Causes, HasSubevent and precedes all express temporal knowledge

Dimensions

	ATOMIC	ConceptNet	WebChild	Other	Wikidata
taxonomic		IsA InstanceOf MannerOf	hasHypernymy	perspective_on (FN) inheritance (FN) hypernym (WN)	subClassOf instanceOf description
part-whole		PartOf HasA MadeOf AtLocation*	physicalPartOf memberOf substanceOf	HasPart (HP) meronym (WN) holonym (WN)	has part member of material used
spatial		AtLocation*	location		location
creation		LocatedNear CreatedBy	spatial		anatomical location creator
utility		ReceivesAction UsedFor CapableOf ¬NotCapableOf	hassynsetmember activity participant	using (FN)	used by use uses
desire/goal	xIntent xWant oWant	CausesDesire MotivatedByGoal Desires ¬NotDesires ObstructedBy			
quality		HasProperty ¬NotHasProperty SymbolOf	shape size color taste_property temperature	frame_element (FN)	color has quality
comparative	xAttr		6.3k relations		
temporal	xNeed xEffect oEffect xReact oReact	HasFirstSubevent HasLastSubevent HasSubevent HasPrerequisite Causes Entails	time emotion prev next	subframe (FN) precedes (FN) inchoative_of (FN) causative_of (FN)	has cause has effect

Dimensions of Commonsense Knowledge

lexical	utility
similarity	desire/goal
distinctness	quality
taxonomic	comparative
part-whole	temporal
spatial	relational-other
creation	

Coverage of dimensions in sources

Dimension	ATOMIC	ConceptNet	WebChild	ROGET	Wikidata-CS	WordNet	FrameNet
lexical		704			0.5	207	14
similarity		255	343	1,023	1	152	0.4
distinctness		22		381	7	4	
taxonomic		244	783		73	89	23
part-whole		19	5,752		8	22	
spatial		28	660		0.5		
creation		0.3			0.2		
utility		69	2,843		2		1
desire/goal	244	20					
quality	143	9	6,510		1		11
comparative			813				
temporal	346	71	2,135		3		0.6
relational-other		1,969	291		6		0.7

Little overlap of knowledge

Sources	part-whole	taxonomic	lexical	distinctness	similarity	quality	utility	creation	temporal	rel-other
CN-RG	- -	- -	- -	4,639 (1.17)	69,353 (5.79)	- -	- -	- -	- -	- -
CN-WD	68 (0.25)	1,888 (0.62)	20 (0.00)	266 (1.00)	102 (0.04)	0 (0.00)	14 (0.02)	0 (0.00)	1 (0.00)	264 (0.01)
CN-WN	4,710 (4.10)	73,123 (15.19)	- -	1,053 (4.65)	19,060 (5.05)	- -	- -	- -	- -	- -
RG-WD	- -	- -	- -	206 (0.05)	127 (0.01)	- -	- -	- -	- -	- -
RG-WN	- -	- -	- -	3,300 (0.87)	71,725 (6.50)	- -	- -	- -	- -	- -
WD-WN	82 (0.07)	1,533 (0.39)	- -	63 (0.62)	26 (0.02)	- -	- -	- -	- -	- -

Zero-shot QA with CSKG dimensions

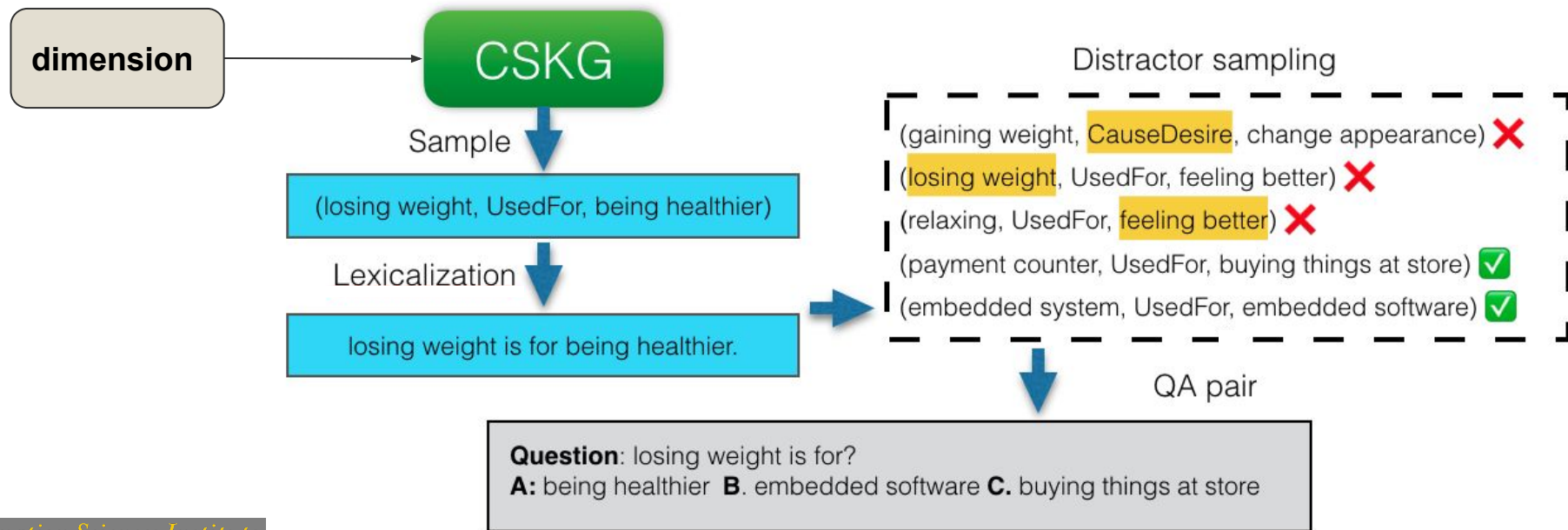
Method

(Ma et al., 2021)

Pretrain LMs with artificial QA sets generated from CSKG

Answer commonsense questions on unseen datasets

One dimension at a time to measure their impact on a task



Slicing CSKG questions along dimensions

Dimensions	Train	Dev
part-whole	87,765	4,620
taxonomic	340,609	17,927
lexical	107,861	5,677
distinctness	20,286	1,068
similarity	166,575	8,768
quality	116,593	12,492
utility	63,862	3,362
creation	304	17
temporal	312,628	31,587
relational-other	242,759	12,777
spatial	21,726	1,144
desire/goal	194,906	20,912

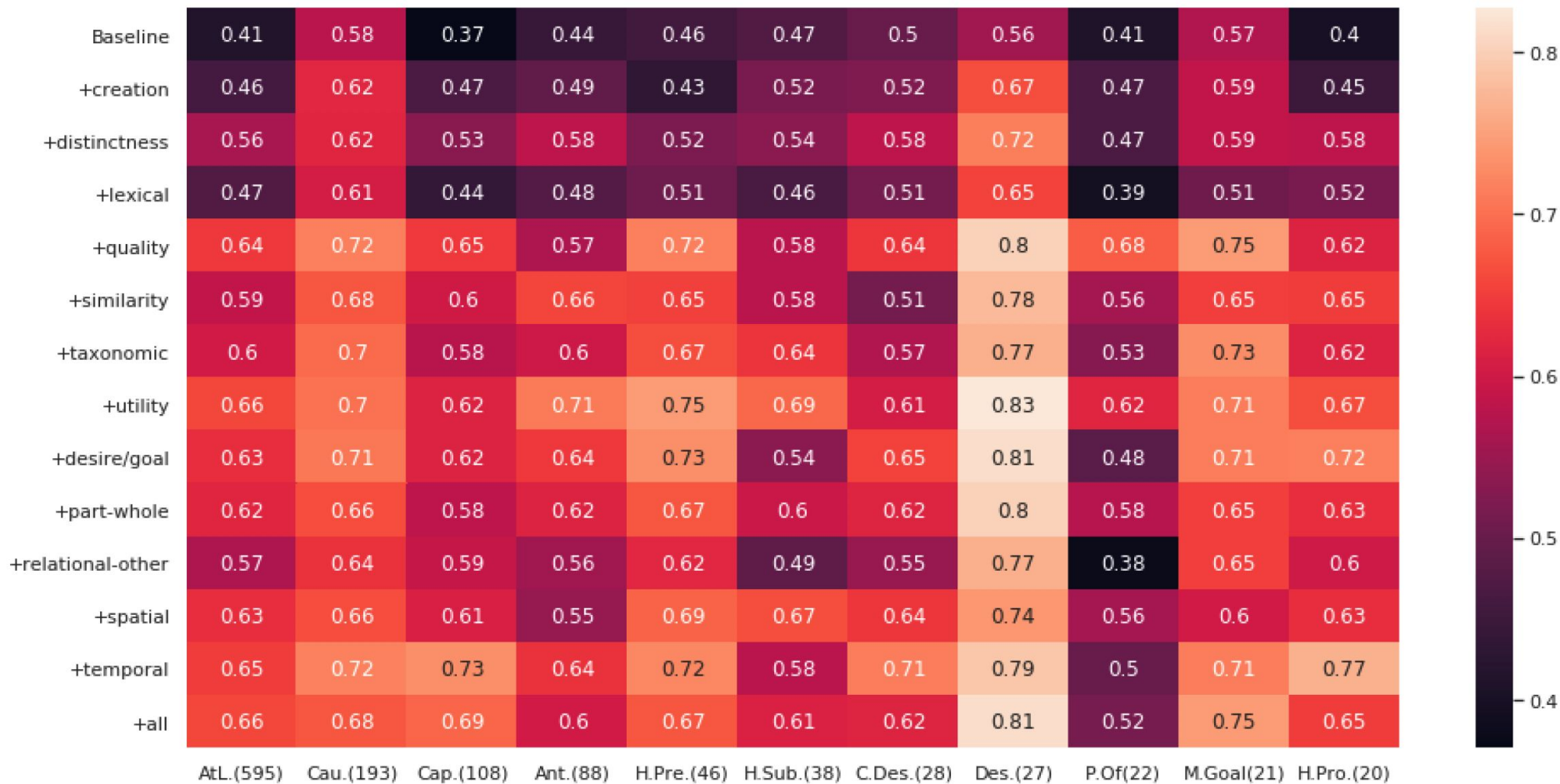
Pre-training language models with dimensions

CSQA = Commonsense QA

SIQA = SocialQA

Dimensions	CSQA	SIQA
Baseline	45.0	47.3
+part-whole	63.0(± 1.4)	52.6(± 1.9)
+taxonomic	62.6(± 1.4)	52.2(± 1.6)
+lexical	49.9(± 2.9)	49.0(± 0.4)
+distinctness	57.2(± 0.5)	50.2(± 1.5)
+similarity	61.4(± 0.8)	53.5(± 0.6)
+quality	65.7(± 0.5)	60.0(± 0.7)
+utility	67.4(± 1.0)	54.8(± 0.7)
+creation	49.9(± 1.1)	47.8(± 0.2)
+temporal	67.3(± 0.3)	62.6(± 0.9)
+relational-other	58.2(± 1.7)	51.3(± 1.7)
+spatial	63.3(± 0.2)	53.1(± 0.3)
+desire/goal	65.0(± 1.8)	60.0(± 0.6)
+all	66.2(± 1.4)	61.0(± 0.7)

Accuracy per question type (CSQA)



Novelty per dimension

Can ‘vanilla’ RoBERTa answer the questions without pretraining?

Dimensions	Dev
part-whole	67.5
taxonomic	57.0
lexical	90.1
distinctness	77.3
similarity	65.6
quality	45.5
utility	67.9
creation	82.4
temporal	47.2
relational-other	37.6
spatial	56.9
desire/goal	48.0

Findings

Consolidation is not trivial but beneficial

- **nodes: improve the graph connectivity**
- **relations: align the knowledge types**

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- **relations: align the knowledge types**

The knowledge dimensions allow us to study the knowledge-task alignment

- **lexical and distinctness knowledge is largely redundant**
- **spatial knowledge is novel but not useful for current tasks**
- **temporal and desire/goal knowledge is both novel and useful**

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- **relations: align the knowledge types**

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We are only starting to understand the coverage/gaps of knowledge sources and evaluation!

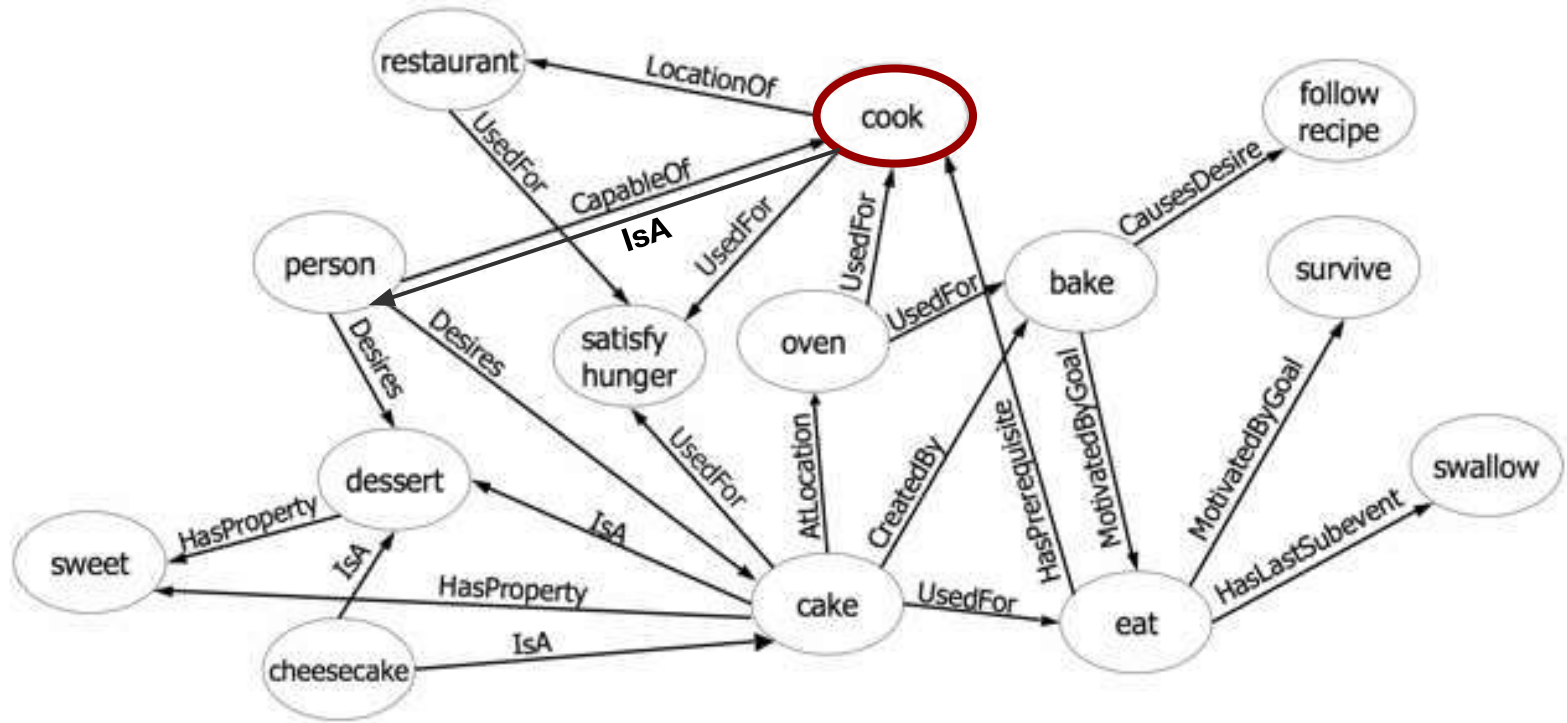
Open challenge: Node resolution

Node label: scene

fn:lu:sensation:scene	FN
fn:lu:locale_by_event:scene	FN
fn:fe:scene	FN
Q7430735	WD
Q67943498	WD
Q16675888	WD
Q1185607	WD
Q282939	WD
/c/en/scene/n/wn/location	CN, WN
/c/en/scene/n/wn/event	CN, WN

/c/en/scene/n/wn/cognition	CN, WN
/c/en/scene/n/wn/artifact	CN, WN
/c/en/scene/n/wn/state	CN, WN
/c/en/scene/n/opencyc/scene_dramatic	CN
/c/en/scene/n/opencyc/image_space	CN
/c/en/picture/n/wn/state	CN, WN
/c/en/scene/n	CN
/c/en/scene	CN
/c/en/scenery/n/wn/artifact	CN, WN

Open challenge: Ambiguity of nodes



Open challenge: Variance of nodes

coffee has...

- en caffeine →
- en caffein →
- en the active ingredient caffeine →
- en caffeine which raises brain reaction time →
- en caffiene →
- en caffiene in it →
- en caffeine →
- en large amounts of caffeine →
- en a slightly bitter taste →
- en a stimulant drug: caffeine →

Open challenge: Relation granularity

ConceptNet

Web Child

/r/HasProperty



-
ability#n#1
age#n#1
appearance#n#1
beauty#n#1
color#n#1
disposition#n#4
emotion#n#1
feeling#n#1
length#n#1
manner#n#1
motion#n#4
personality#n#1
physical_property#n#1

quality#n#1
sensitivity#n#2
shape#n#2
size#n#1
sound#n#1
state#n#2
strength#n#1
structure#n#2
sustainability#n#1
tactile_property#n#1
taste_property#n#1
temperature#n#1
trait#n#1
weight#n#1

Open challenge: Knowledge filtering



An English term in ConceptNet 5.8

Sources: DBPedia 2015, JMDict 1.07, English Wiktionary, and Open Multilingual WordNet
[View this term in the API](#)

[Documentation](#)

[FAQ](#)

[Chat](#)

[Blog](#)

Synonyms

fi vesikuolio (n, state) →
fr noma (n, state) →
id noma (n, state) →
ja 水瘡 (n) →
ja 野間 (n, state) →
pt Noma (n, state) →
ja 水瘡 →

Related terms

en destruction →
en disease →
en gangrenous →
en tissue →
en noma pudendi (n) →
ja 水瘡 (n) →
pt noma (n) →

Etymological roots of "noma"

grc νομή →
la noma →
la nome →

noma is a type of...

en ulcer (n, state) →
en disease (n) →
en restaurant (n) →

Open challenge: Missing facts

Location of barbecue

- en australia →
- en Australia →
- en Detroit →
- en a garage →
- en a Michigan park →
- en a retail store →

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