

# Agenda

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

# 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<sub>A</sub> makes Something<sub>A</sub> (that is food) Causes/Enables Someone<sub>A</sub> eats Something<sub>A</sub></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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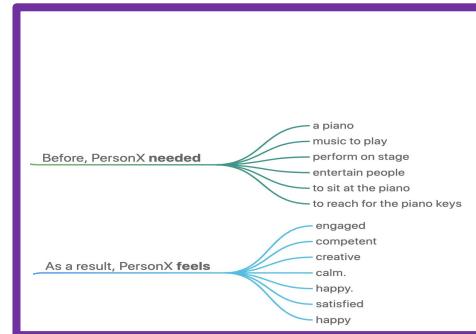
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ATOMIC: to play piano, a person needs to sit at it, on stage and reach for the keys; feelings

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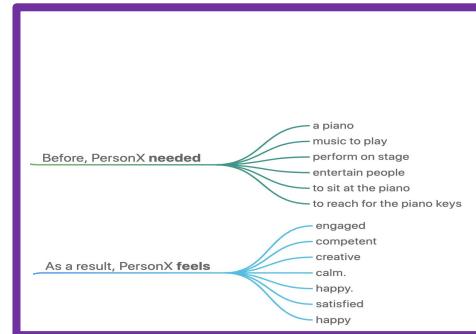
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FrameNet:  
performer entertains audience

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 Performer generates the Performance which the Audience perceives.

Performer [Perfer]

The Performer provides an experience for the Audience.

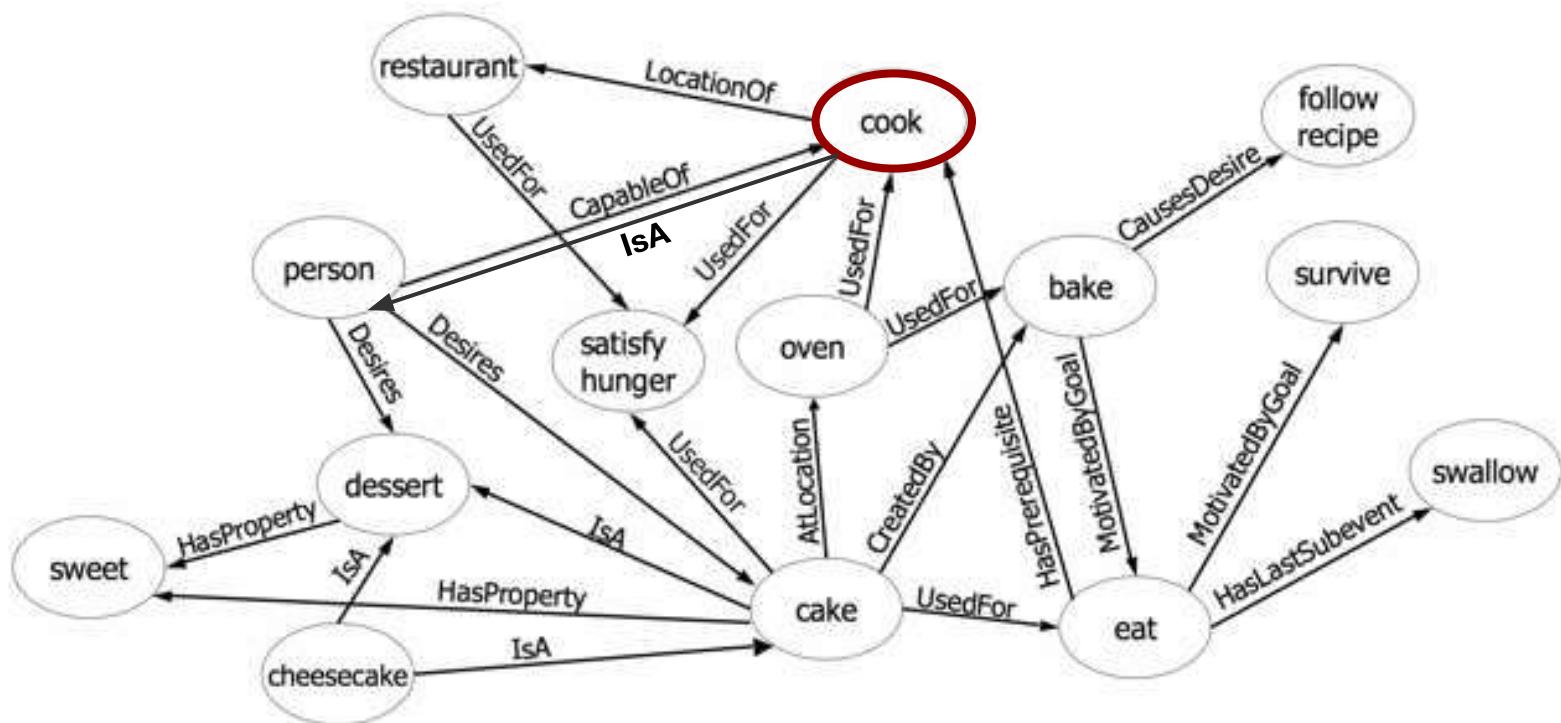
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# Challenge: Knowledge granularity

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

# Challenge: Imprecise descriptions



# Challenge: Sparse overlap and mappings

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

	<b>mappings</b>
<b>ConceptNet</b>	WordNet, DBpedia, OpenCyc, Wiktionary
<b>WebChild</b>	WordNet
<b>ATOMIC</b>	ConceptNet, Cyc
<b>Wikidata</b>	various
<b>CEO</b>	FrameNet, SUMO
<b>WordNet</b>	
<b>Roget</b>	
<b>VerbNet</b>	FrameNet, WordNet
<b>FrameNet</b>	
<b>Visual Genome</b>	WordNet

# Challenge: Modeling of relations

ConceptNet

/r/HasProperty



Web Child

-  
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

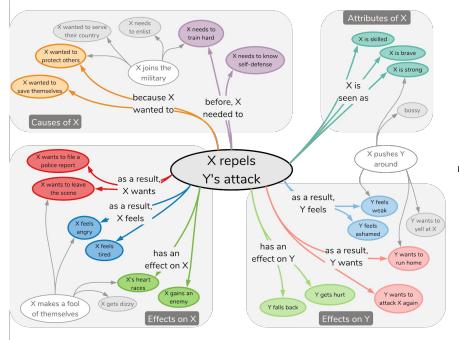
# Overview of node mappings

251,517 mw:SameAs

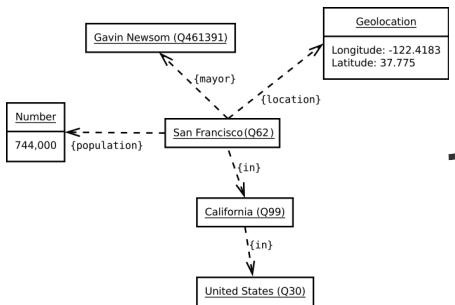
45,659 mw:HasInstance

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



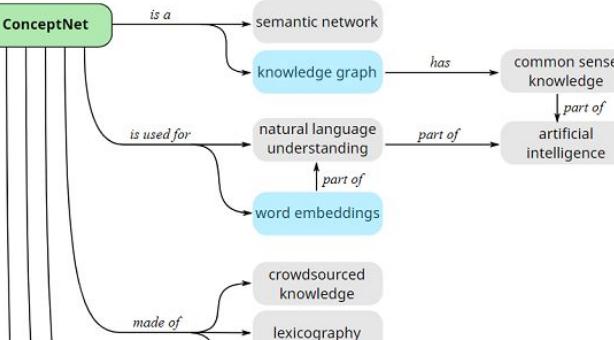
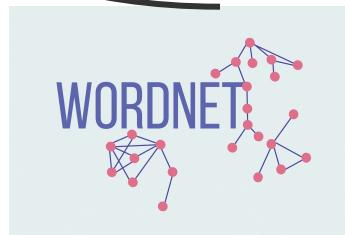
ATOMIC (Sap et al. 2019)



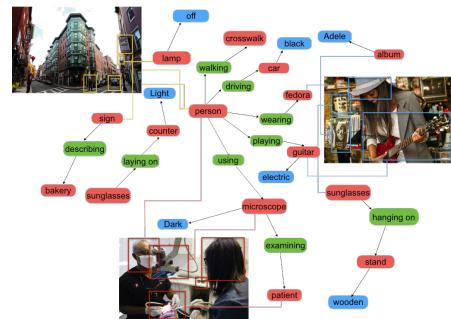
Wikidata (Vrandecic and Krotzsch 2014)

WordNet (Miller 1995)

CSKG (Ilievski et al. 2020)

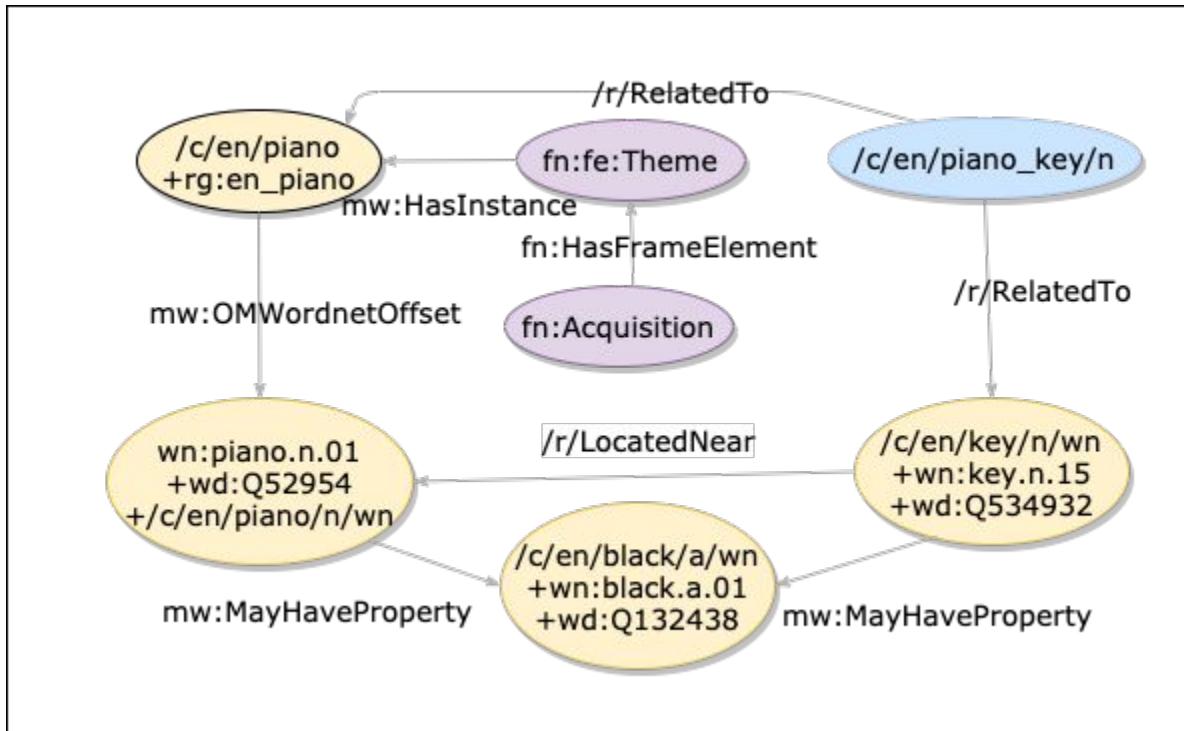


ConceptNet (Speer, Chin and Havasi 2017)

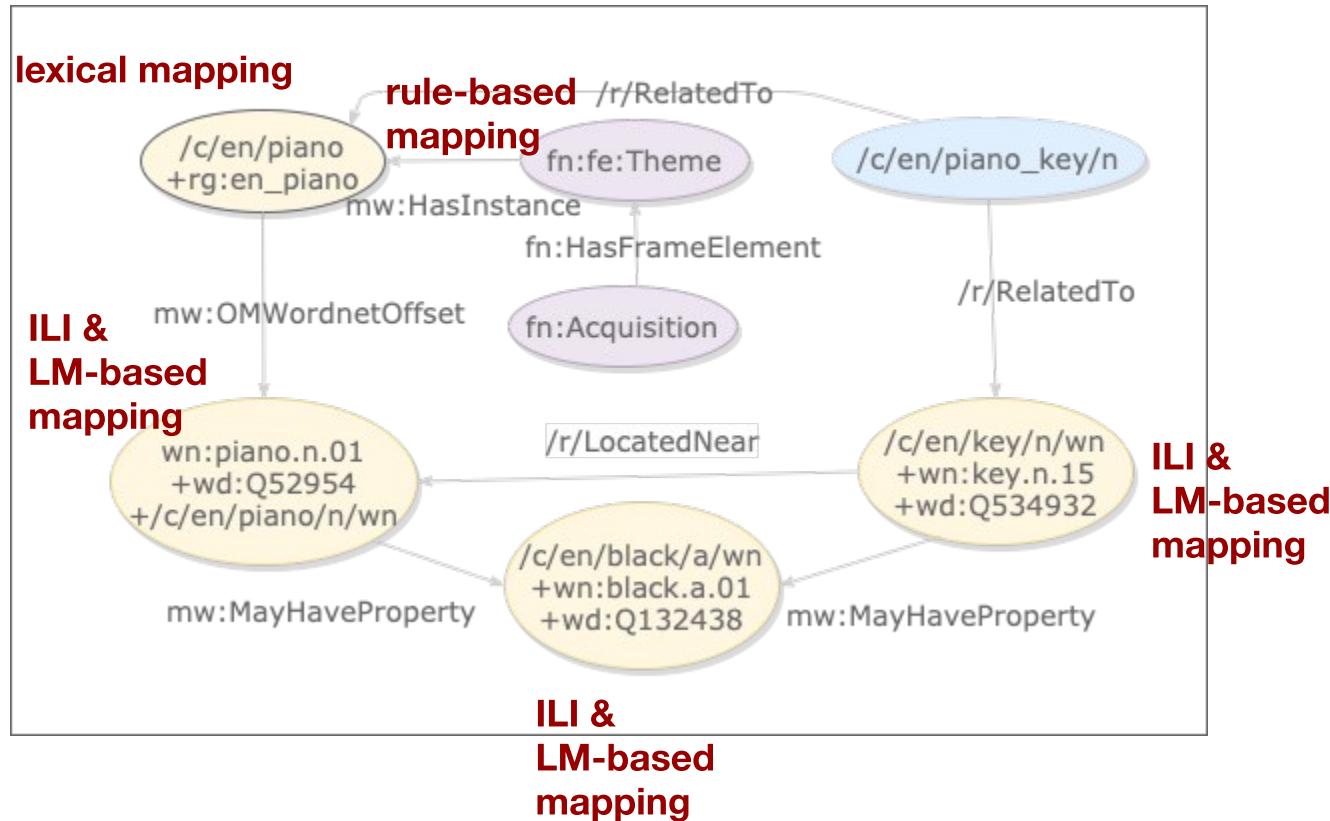


Visual Genome (Krishna et al. 2017)

# 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

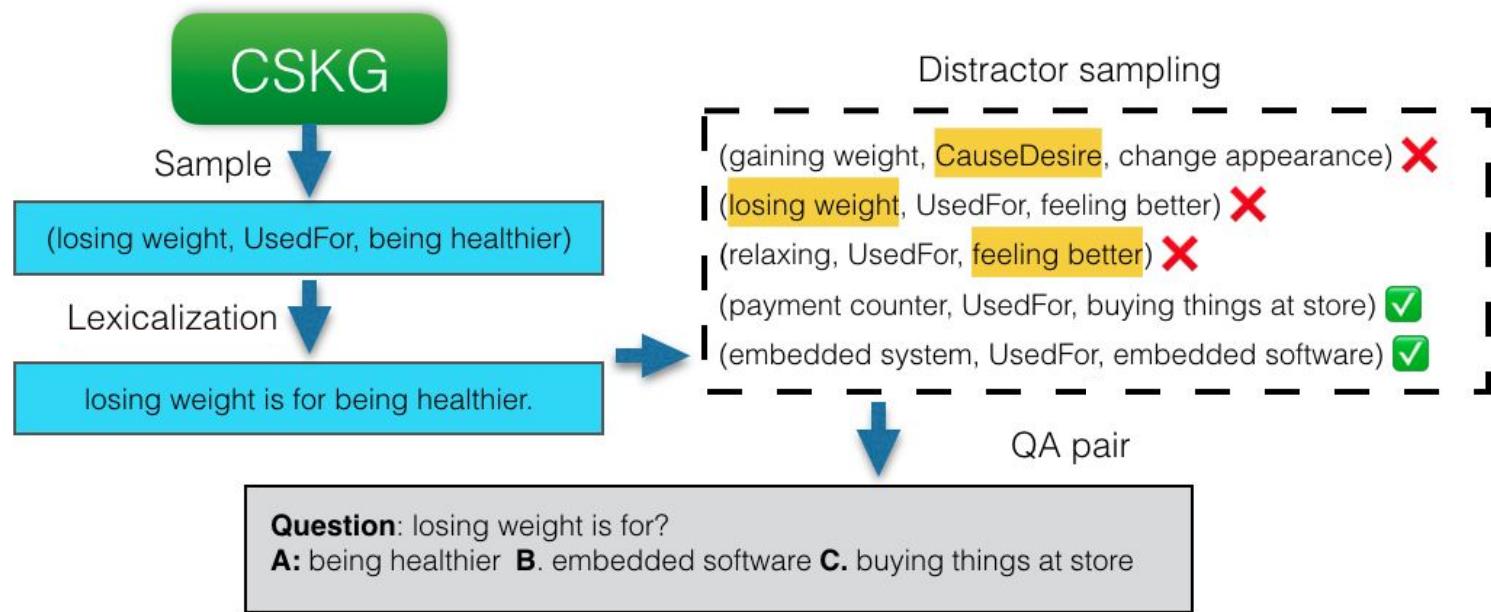
1. /c/en/**chemical\_compound/n**
2. /c/en/**change/n/wn/artifact**
3. /c/en/**natural\_science/n/wn/cognition**
4. /c/en/**chromatic/a/wn**
5. /c/en/**organic\_compound**

# Zero-shot QA with CSKG

**Method**  
(Ma et al., 2021)

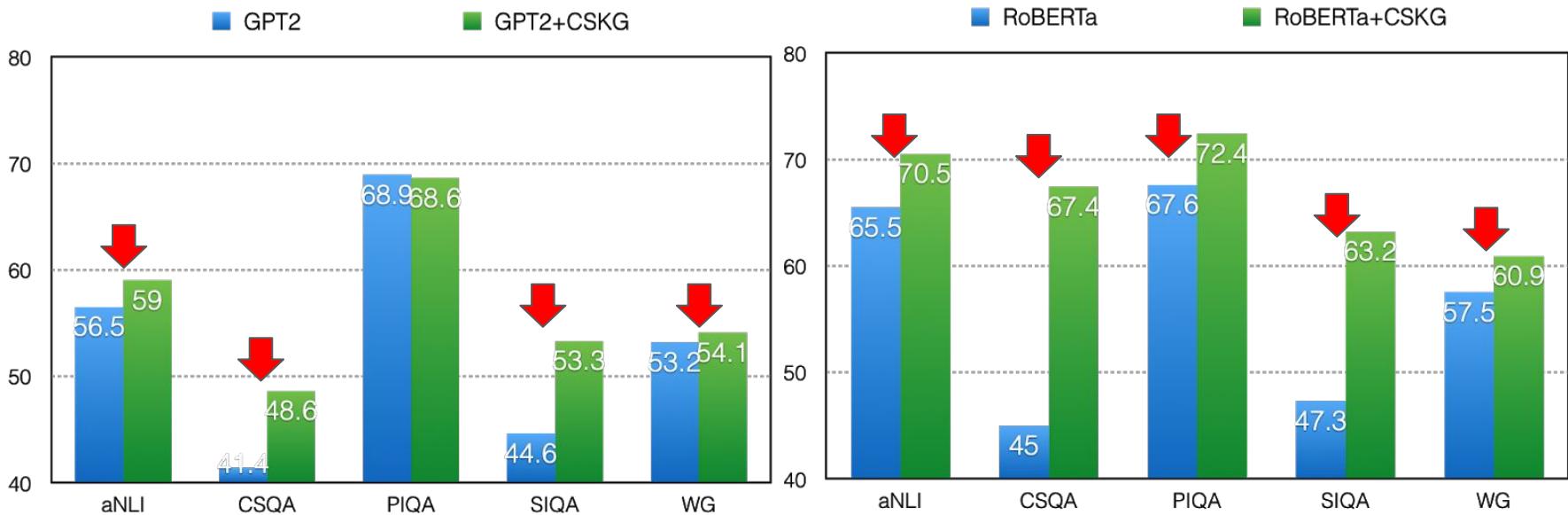
Pretrain LMs with artificial QA sets generated from CSKG

Answer commonsense questions on unseen datasets



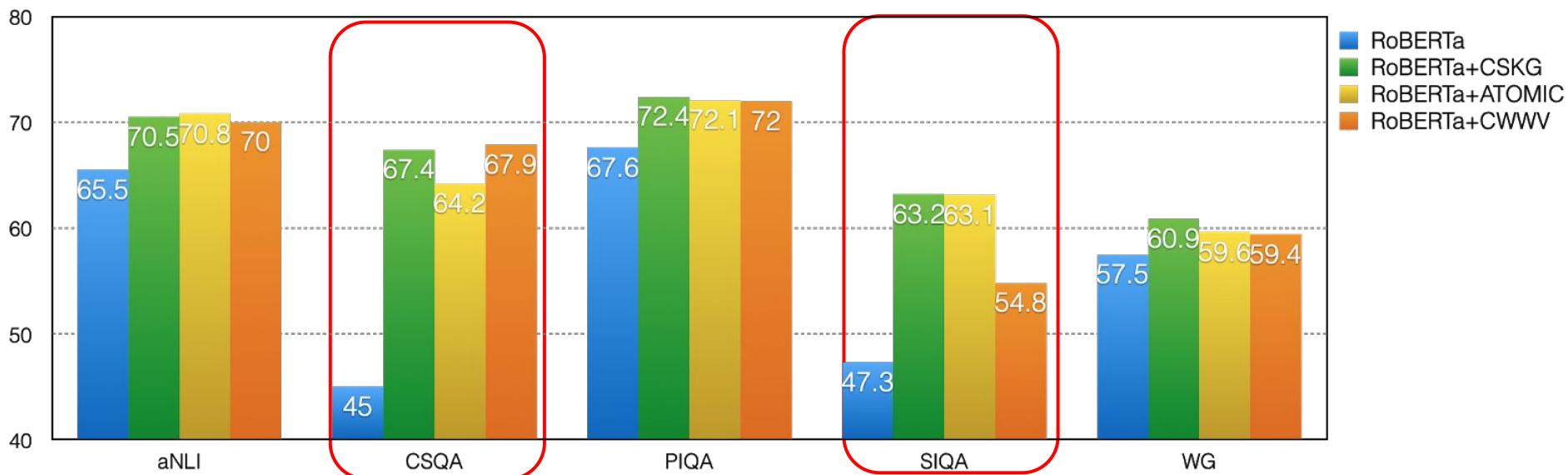
# Main Results

- 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	xAttr	HasProperty ¬NotHasProperty SymbolOf	shape size color taste_property temperature	frame_element (FN)	color has quality
comparative			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 similarity		704			0.5	207	14
distinctness		255	343	1,023	1	152	0.4
taxonomic		22		381	7	4	
part-whole		244	783		73	89	23
spatial		19	5,752		8	22	
creation		0.3			0.5		
utility		28	660		0.2		
desire/goal	244	69	2,843		2		1
quality	143	20					
comparative		143	9	6,510	1		11
temporal				813			
relational-other	346	71	2,135		3		0.6
		1,969	291		6		0.7

# Little overlap of knowledge

Sources	part-whole	taxonomic	lexical	distinctness	similarity	quality	utility	creation	temporal	rel-other
<b>CN-RG</b>	-	-	-	4,639	69,353	-	-	-	-	-
	-	-	-	(1.17)	(5.79)	-	-	-	-	-
<b>CN-WD</b>	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)
<b>CN-WN</b>	4,710 (4.10)	73,123 (15.19)	-	1,053 (4.65)	19,060 (5.05)	-	-	-	-	-
<b>RG-WD</b>	-	-	-	206	127	-	-	-	-	-
	-	-	-	(0.05)	(0.01)	-	-	-	-	-
<b>RG-WN</b>	-	-	-	3,300	71,725	-	-	-	-	-
	-	-	-	(0.87)	(6.50)	-	-	-	-	-
<b>WD-WN</b>	82 (0.07)	1,533 (0.39)	-	63 (0.62)	26 (0.02)	-	-	-	-	-

# Zero-shot QA with CSKG dimensions

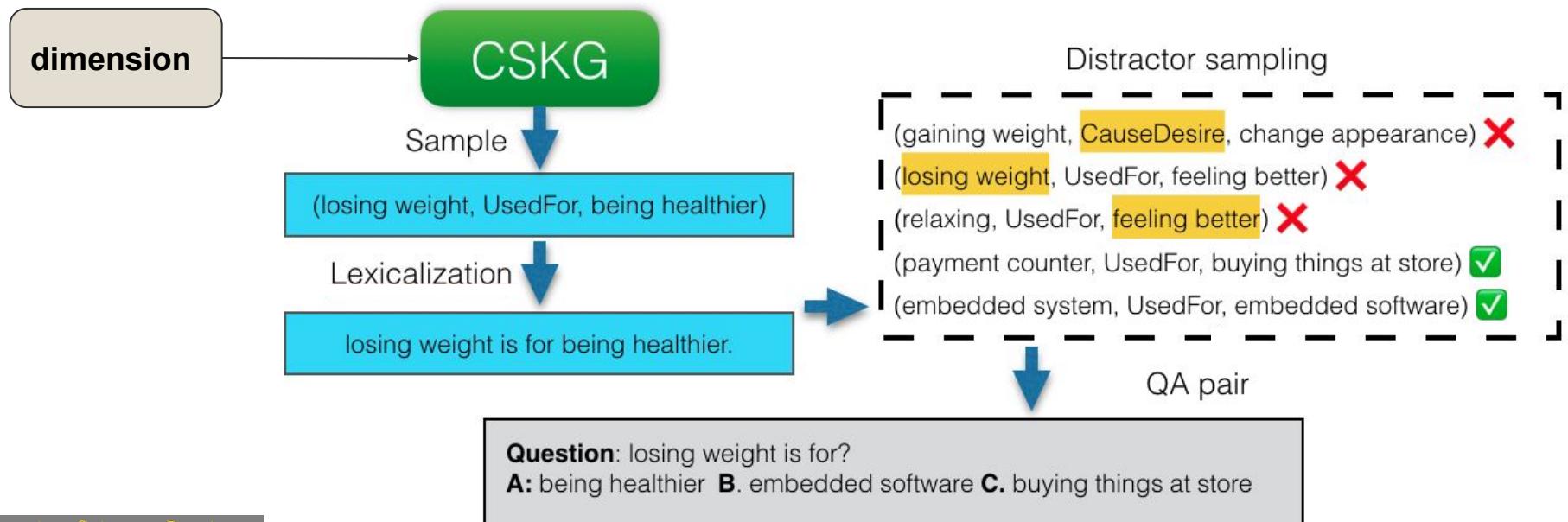
## Method

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

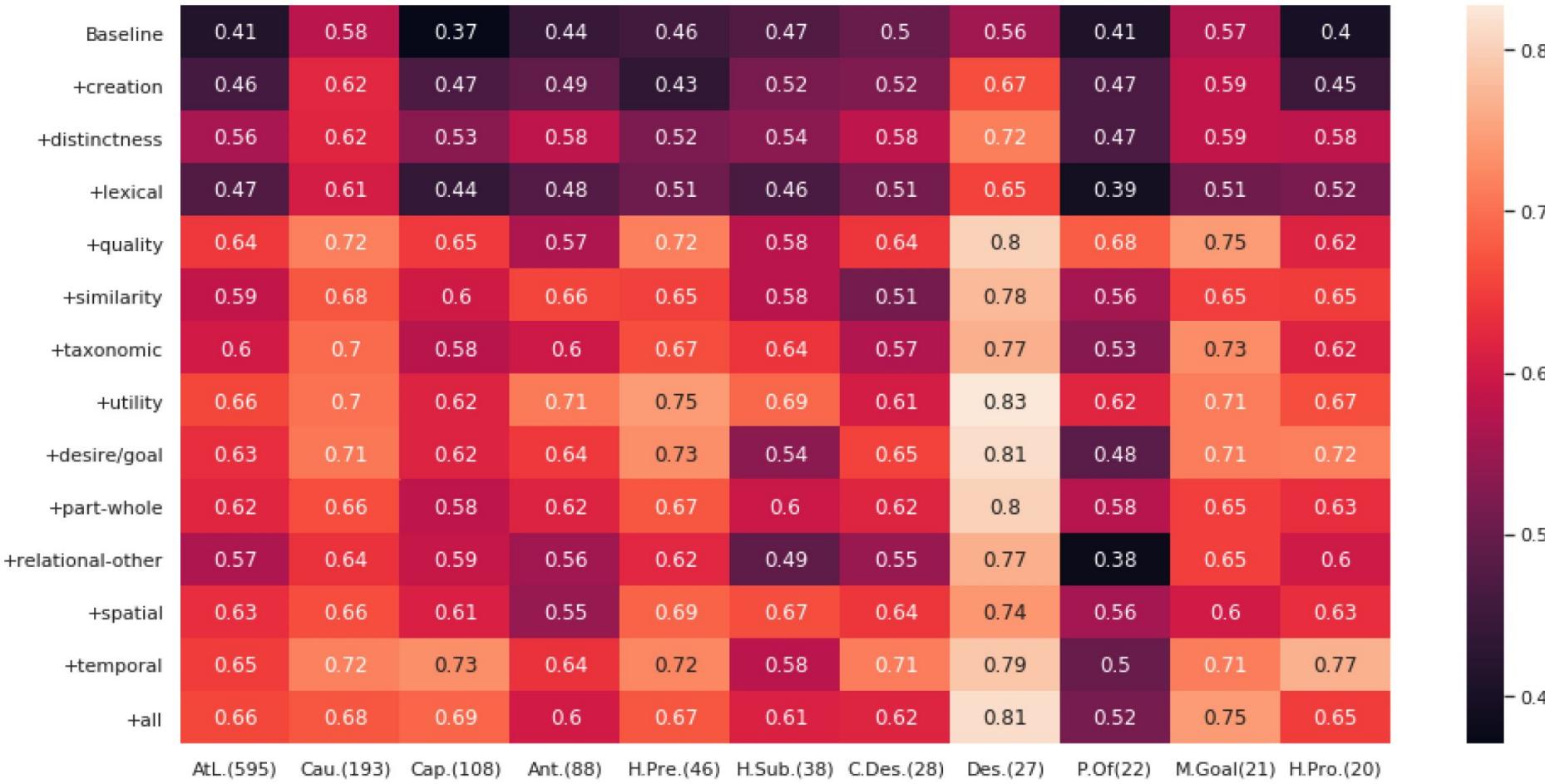
# Pre-training language models with dimensions

CSQA = Commonsense QA

SIQA = SocialIQA

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

# 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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**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**

# Findings

**Consolidation is not trivial but beneficial**

- **nodes: improve the graph connectivity**
- **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**

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

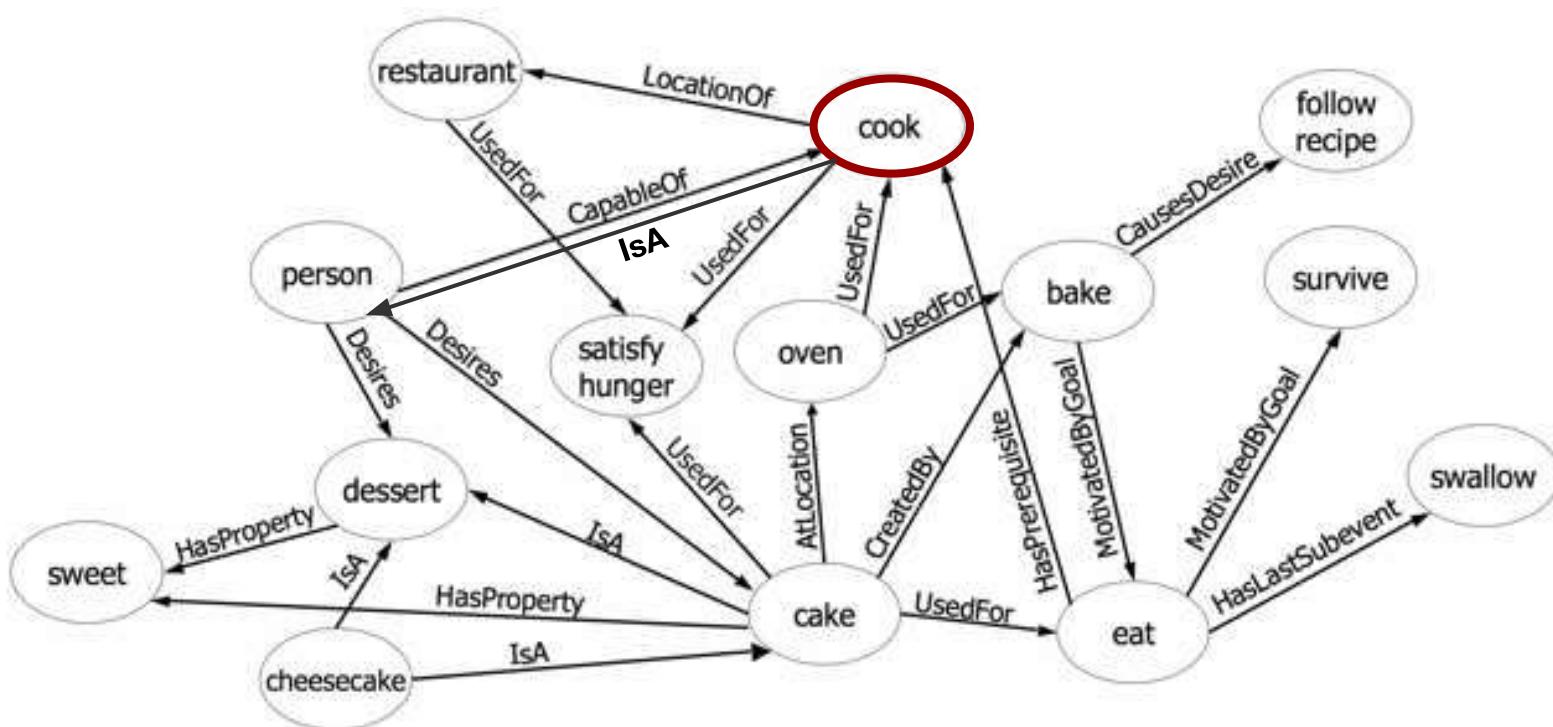
# Open challenge: Node resolution

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

# Open challenge: Relation granularity

ConceptNet

/r/HasProperty



Web Child

- 
- 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 水癌 (n) →
- ja 野間 (n, state) →
- pt Noma (n, state) →
- ja 水癌 →

## Related terms

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

## Etymological roots of "noma"

- grc νομα →
- 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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