Experiments in Learning to Solve Formal Analogical Equations

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Analogy Formal Analogy

Analogy

[x : y :: z : t] stands for a proportional analogy

x is to y as z is to t



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Ph. Langlais

LEARNING TO SOLVE FORMAL ANALOGICAL EQUATIONS

Analogy Formal Analogy

Formal Analogies (or Analogies on Forms)

• $[x \setminus y \bigcirc z \supset t]$ denotes a formal analogy

- $1 \quad [funny \setminus funniest \bigcirc lucky \supset luckiest]$
- 2 [suggère \ suggérer ingère ⊃ ingérer]
- 3 [This guy drinks too much \ This boat sinks () These guys drank too much ⊃ These boats sank]
- Several operational definitions of formal analogy [?, ?]

Analogy Formal Analogy

Definition of [?]

 $[x \setminus y \bigcirc z \supset t]$ iff we can find *d*-factorizations f_x, f_y, f_z and f_t such that, $\forall i \in [1, d]$:

$$(f_{\mathbf{y}}^{(i)}, f_{\mathbf{z}}^{(i)}) \in \left\{ (f_{\mathbf{x}}^{(i)}, f_{\mathbf{t}}^{(i)}), (f_{\mathbf{t}}^{(i)}, f_{\mathbf{x}}^{(i)}) \right\}$$

[this guy drinks too much \ this boat sinks ○ these guys drank too much ⊃ these boats sank] because:

Х	\equiv	this	guy	ϵ	dr	inks	too much
у	≡	this	boat	ϵ	s	inks	ϵ
z	\equiv	these	guy	s	dr	ank	too much
+		Also a second	la a a t	_	_	a se la	

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Finite State Automaton alea Word2Vec Our Approach

(Yvon & al., 2004)
$$[x \setminus y \bigcirc z \supset t] \iff t \in (y \bullet z) \setminus x$$

shuffle a • b read sequences in a and b from left to right, allowing to switch from one string to another

spondy on dontil algiatis et ond $spondony la ltitisgia \in spondy la lgia \circ on dontitis$

• complement $a \setminus b$ strings obtained by removing substring b in a

 $spondylitis \in spondyondontilalgiatis \setminus ondontalgia$ $spydoniltis \in spondyondontilalgiatis \setminus ondontalgia$

 $\{(y \bullet z) \setminus x\}$ is a rational language

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Solver alea [?]

- building the automaton typically is combinatorial
- ▶ alea randomly samples p shuffles of y and z and complements them (with x)

notes:

- ▶ for too low values of *p*, alea may fail to identify a solution
- different shuffles may lead (after complementation) to the same solution
- solutions are ranked in decreasing order of frequency

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[this guy drinks too much \setminus this boat sinks \bigcirc those guys drink too much \supset ?]

$\rho = 20$	n = 8			
t = 0.0003	$rank=\phi$			
thos_boate_sinks (2)				
tho_boatse_sinks (2)				
thoboatse	<i>sinks</i> (2)			

 $\begin{array}{ll} \rho = 1000 & n = 28 \\ t = 0.009 & rank = 2 \\ \hline those_boat_ssink \ (5) \\ those_boats_sink \ (5) \\ thoes_tboa_sinks \ (5) \end{array}$

- ρ: nb of shuffles sampled
- n: # of solutions proposed

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 $ho = 10^6$ n = 19796 t = 3.82 rank = 10those_boat_sinks (2550) those_boat_sink (1037) those_boat_ssink (999)

- rank: rank of the reference solution (φ ≡ not found)
- ► *t*: time (in sec.)

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Solver word2vec

•
$$queen - king \sim woman - man$$

$$\hat{t} = \operatorname*{argmax}_{t \in V} \cos(t, z - x + y)$$

- Can also solve syntactic equations: [work \ worked ○ accept ⊃ ?] ▷ accepted
- Note: rank words in V but does not generate new ones

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Learning to Solve Equations

- ▶ making use of a training set of equations and their solutions $T = \{((x, y, z), t)\}$ where $[x \setminus y \bigcirc z \supset t]$
- using structured learning
 - given g : I³ × I → ℝ which evaluates a fit between a triple of strings in I, i ≡ (x, y, z), and any string t, we seek to find (search):

$$\hat{t} = \mathop{\mathrm{argmax}}_{t \in \mathcal{I}} g(i,t)$$

- we assume a linear model for $g = \langle \mathbf{w}, \Phi(i, t) \rangle$ parametrized by a feature vector \mathbf{w} in \mathbb{R}^{K} and a feature map $\Phi(i, t)$ decomposed into K binary feature fonctions $\phi_k : (i, t) \to \{0, 1\}$.
- we wish to adjust w so as to minimize over T the number of search errors, thanks to the voted perceptron algorithm [?]

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Search



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Peculiarities of the Search

- ► Tree of depth |x| + |y| + |z| with a branching factor close to 2 on average (too many nodes)
- Only few actions increase the prefix of a solution (in particular, many states with an empty prefix)

In practice, we:

- search space organized as a graph
- check that complementation is still possible
- ► control the maximum number of *X* or *Y* actions that can take place without complementing with *x*

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Feature Map

3 families of binary features for characterizing $\langle s, i, j, k, p \rangle$:

language model (14 features) evaluating the likelihood of the prefix p (or shuffle s) so far, according to a n-gram LM trained on an out-domain monolingual corpus.

 $\blacktriangleright \exists ?i: p_{LM}(p_i|p_{i-2}p_{i-1}) < \delta$

edit-distance (20 features) to enforce that the solution shares with y and z subsequences

[reader : unreadable :: doer : undoable]

search-based (20k features)

- global: e.g. number of Y or Z actions in a row
- ► local: e.g. $(x_i, y_j, z_k) \equiv (\mathbf{r}, \mathbf{a}, \mathbf{d})$

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Averaged Voted Perceptron [?]

```
\mathbf{w}, \mathbf{w}_a \leftarrow \mathbf{0}
e \leftarrow 0
repeat
    e \leftarrow e + 1
     for all example (i, t) \in D do
         \hat{t} = \operatorname{argmax}_{t} \mathbf{w}^{T} \Phi(i, t)
         if \hat{t} \neq t then
              \mathbf{w} \leftarrow \mathbf{w} + \Phi(i, t) - \Phi(i, \hat{t})
              \mathbf{w}_a \leftarrow \mathbf{w}_a + \mathbf{w}
until converged
return \mathbf{w}_a/e.|\mathcal{D}|
```

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Averaged Voted Perceptron Variants

- Not suited for inexact search (our case) [?]
 - update only if $\Phi(i,t) \Phi(i,\hat{t}) < 0$
- We implemented 4 different variants (using forced-decoding)



standard do not care

safe remove the example if condition is not satisfied

early update on a prefix whenever there is no prefix of a reference solution in the beam

late update on the largest prefix that satisfies the condition



Metrics Solving Analogies on Words Solving Analogies on Phrases

Metrics

accuracy percentage of test equations for which the solver output the correct solution at rank 1 silence percentage of test equations without any solution proposed

Metrics Solving Analogies on Words Solving Analogies on Phrases

Dataset google

Formal analogies of the dataset released by [?]

of analogies: 4977 word's avr. length: 7

adjective-adverbeADJ-ADV		$[amazing \setminus amazingly \bigcirc serious \supset seriously]$
opposite	OPP	$[certain \setminus uncertain \bigcirc competitive \supset uncompetitive]$
comparative	COMP	$[fast \setminus faster \bigcirc bright \supset brighter]$
superlative	SUP	$[warm \setminus warmest \bigcirc strange \supset strangest]$
present-participle	PP	$[code \setminus coding \bigcirc dance \supset dancing]$
nationality-adverb	NAT	$[Australia \setminus Australian \bigcirc Croatia \supset Croatian]$
past-tense	PAST	$[decreasing \setminus decreased \bigcirc listening \supset listened]$
plural	PLUR	$[eye \setminus eyes \bigcirc donkey \supset donkeys]$
plural-verbs	PL-VB	$[listen \setminus listens \bigcirc eat \supset eats]$

Metrics Solving Analogies on Words Solving Analogies on Phrases

Dataset: msr

Formal analogies of the dataset released by [?]

# of analogies: 3664	word's avr. length: 6
JJ-JJR	$[\mathit{high} \setminus \mathit{higher} \bigcirc \mathit{wild} \supset \mathit{wilder}]$
JJR-JJ	$[greater \setminus greatest \bigcirc earlier \supset earliest]$
JJS-JJ	$[\mathit{low} \setminus \mathit{lowest} \bigcirc \mathit{short} \supset \mathit{shortest}]$
NN-NNPOS	[problem \ problems \bigcirc program \supset programs]
VB-VBP	$[take \setminus takes \bigcirc run \supset runs]$
VB-VBD	[prevent \ prevented \bigcirc consider \supset considered]
NNPOS-NN	$[days \setminus day \bigcirc citizens \supset citizen]$
VBZ-VBD	$[believes \setminus believed \bigcirc likes \supset liked]$

Metrics Solving Analogies on Words Solving Analogies on Phrases

Accuracy as a function of the beam size (or ρ)

training on msr and testing on google (left), or the other way round (right)



Metrics Solving Analogies on Words Solving Analogies on Phrases

A Comparison to State-of-the-art

trust early if the solution defines with the equation a formal analogy, word2vec otherwise

	msr	google
word2vec	67%	63%
early+word2vec	72%	71%
Levy et al., 2015	72.9%	75.8%

Metrics Solving Analogies on Words Solving Analogies on Phrases

Comparison by Types



Metrics Solving Analogies on Words Solving Analogies on Phrases

Dataset

- Much harder to get. We did this:
 - 1 Train a phrasal translation table (from an EN-ES parallel corpus)

a actualizar los acuerdos | to update the agreements | 0.0047 a cambiar la base | to change the basis | 0.0035545 basado en el trabajo de | based on the efforts of | $2.02579e^-$

- 2 Split good enough associations into R a reference set, and M a translation memory
- 3 For each $(s, e) \in R$,
 - use an analogical device to translate *s* into *e*, using *M* as a memory, that is, finding $(x, x'), (y, y'), (z, z') \in M$ such that: $[x \setminus y \bigcirc z \supset s]$ and $[x' \setminus y' \bigcirc z' \supset e]$
 - collecting $[x' \setminus y' \bigcirc z' \supset e]$ (analogies in the English side)

4 Select:

- simple: a random subset of those equations
- hard: a sampling of equations more difficult for alea to solve

Metrics Solving Analogies on Words Solving Analogies on Phrases

Dataset: simple

S	simple phrase's avr. length: 16				
[international investigation an international investigation	\ ⊃	international democracy an international democracy	0]
[young girls young girls and	\ ⊃	training of young girls training of young girls and	0]
[political situation is viable the political situation is viable	\ ⊃	political situation is still the political situation is still	0]

Metrics Solving Analogies on Words Solving Analogies on Phrases

Dataset: hard

h	ard phra	ase's	avr. length: 17		
[adopted recently by study published recently by	\ ⊃	recently adopted by study recently published by	0]
[competition and the competition and of the	\ ⊃	competition and against of competition and against	0]
[their governments to their governments and to	\ ⊃	their governments are and their governments are	0]

Metrics Solving Analogies on Words Solving Analogies on Phrases

Accuracy (Silence)

test \equiv simple Structured solvers trained on 10 epochs with $\eta=$ 7, against alea with $\rho=1000$

train	simple	hard
standard	30.6 (7)	30.1 (10)
early	38.4 (8)	27.6 (10)
late	26.9 (9)	20.3 (11)
safe	25.4 (8)	25.6 (13)
alea	33	(0)

- alea is competitive, but outperformed by early
- preferable to train on simple analogies

Metrics Solving Analogies on Words Solving Analogies on Phrases

Accuracy (Silence)

test \equiv hard Structured solvers trained on 10 epochs with $\eta=$ 7, against alea with $\rho=1000$

train	simple	hard
standard	19.6 (17)	24.0 (56)
early	26.0 (6)	22.9 (7)
late	18.9 (9)	20.1 (6.7)
safe	14.6 (18.7)	21.0 (56)
alea	18 (0)	

- more challenging dataset !
- structured learning systematically better
- preferable to train on harder to solve analogies
- silence rate rather high

Metrics Solving Analogies on Words Solving Analogies on Phrases

Accuracy on hard

- after 1 and 10 epochs
- for 2 values of η: 7 and 20



- more epochs is (of course) preferable
- opening the search space as well (expectedly)
- early seems overall the best variant tested

Conclusion

Futur

En cours

probabiliser le solveur (thèse de Rafik Rhouma)

Long terme

- probabiliser tout le processus (search, solveur, agrégation)
- passage à l'échelle
- noyau analogique ?
- analogie sur les arbres [?, ?]

Bibliography I