



Probabilistic Refinement Algorithms for the Generation of Referring Expressions



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INTRODUCTION

We propose a **refinement** algorithm that generates referring expressions that are:

- Relational: *ball on the blue cube*
- Overspecified: *small green ball on the blue cube*
- Non-deterministic: *green ball, small ball on the right*

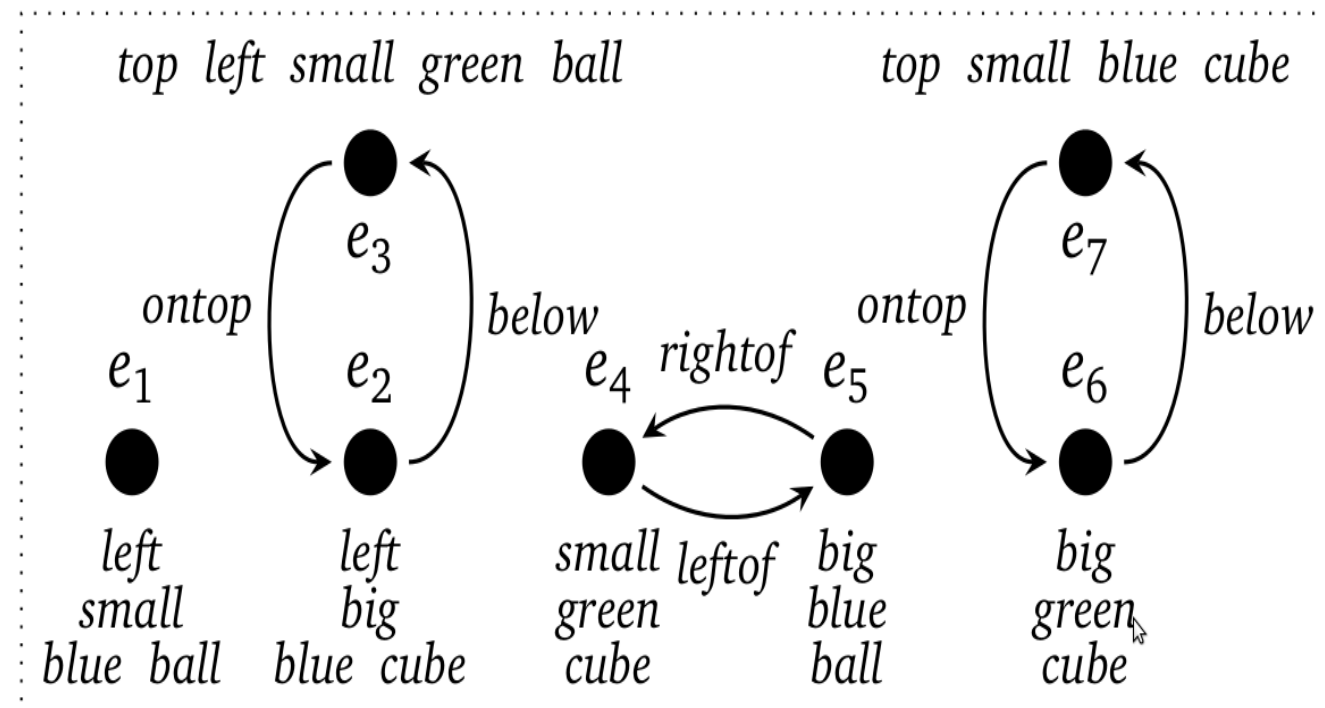
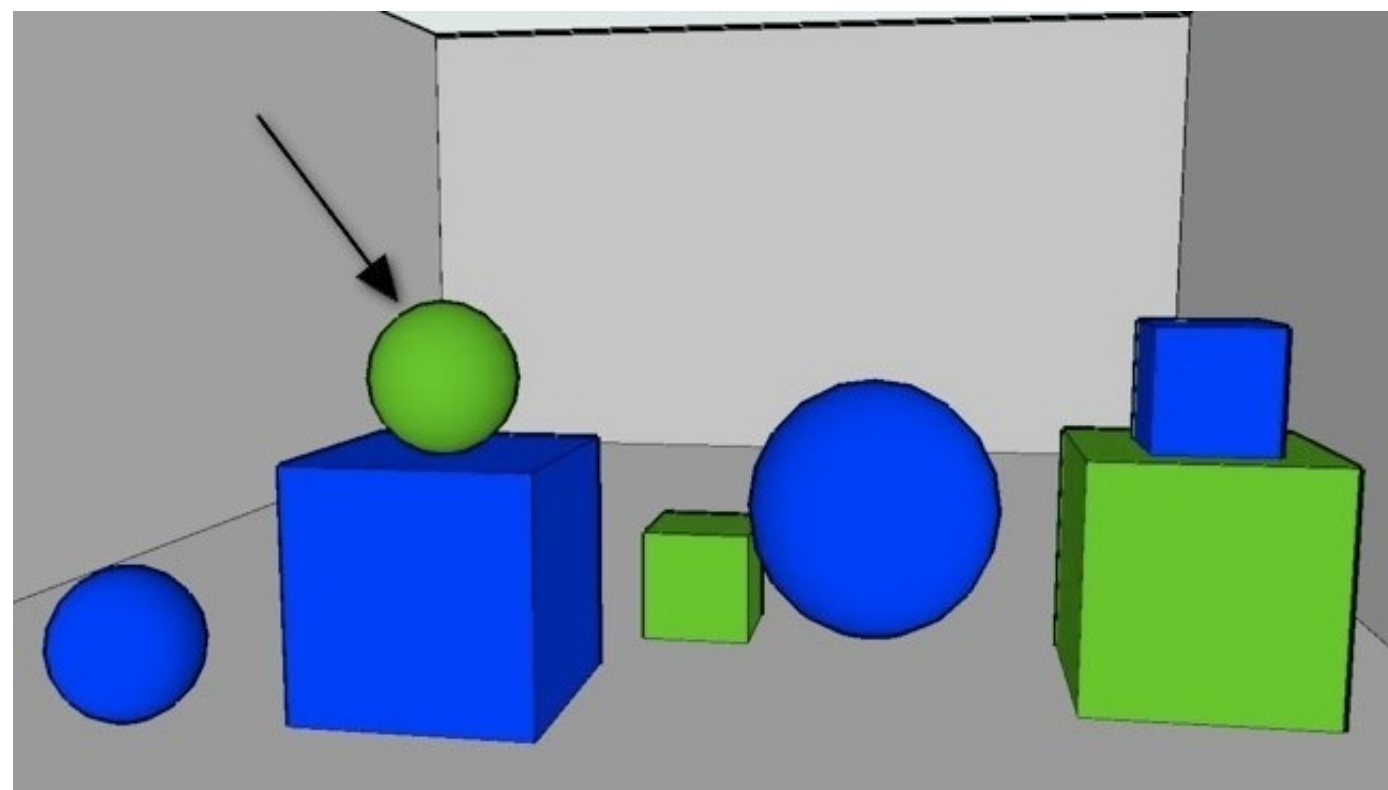


Figure 2: Scene as a relational model

LEARNING PROBABILITIES OF USE

If there is corpus available:

$$R.p_{use} = \frac{\#appear(R)}{\#R_e}$$

Feature used to learn $R.p_{use}$ when there is not corpus available

R.target-has	true if the target is in R
R.landmark-has	true if a landmark is in R
R.discrimination	1 / objects in the model that have R
#bin-relations	number of binary relations of the target
#relations	number of relations of the target

REFINEMENT ALGORITHMS

NON-DETERMINISM AND OVER-SPECIFICATION

Algorithm 1: Computing \mathcal{L} -similarity classes

```

RE ← {T}
repeat
  while  $\exists(\varphi \in RE).(\|\varphi\| > 1)$  do
    RE' ← RE
    for  $(R, R.p_{use}) \in Rs$  do
      if  $R.rnd_{use} \leq R.p_{use}$  then
        for  $\varphi \in RE$  do add $_{\mathcal{L}}$ (R,  $\varphi$ , RE)
        if RE ≠ RE' then
          exit
    if RE = RE' then
      exit
  for  $(R, R.p_{use}) \in Rs$  do  $R.p_{use} \leftarrow R.p_{use} + R.inc_{use}$ 
until  $\forall (R, R.p_{use}) \in Rs. (R.p_{use} \geq 1)$ 

```

Probabilistic
Non-deterministic
step

Completeness
step

Algorithm 2: add $_{\mathcal{L}}$ (R, φ , RE)

```

if FirstLoop? then
  Informative ← TRUE
else Informative ←  $\|\psi \sqcap \exists R.\varphi\| \neq \|\psi\|$ ;
for  $\psi \in RE$  with  $\|\psi\| > 1$  do
  if  $\psi \sqcap \exists R.\varphi$  is not subsumed in RE and
   $\|\psi \sqcap \exists R.\varphi\| \neq \emptyset$  and
  Informative then
    add  $\psi \sqcap \exists R.\varphi$  to RE
    remove subsumed formulas from RE

```

Overspecification
Step/ (Egocentric)

Refinement
Step/(Adjustment)

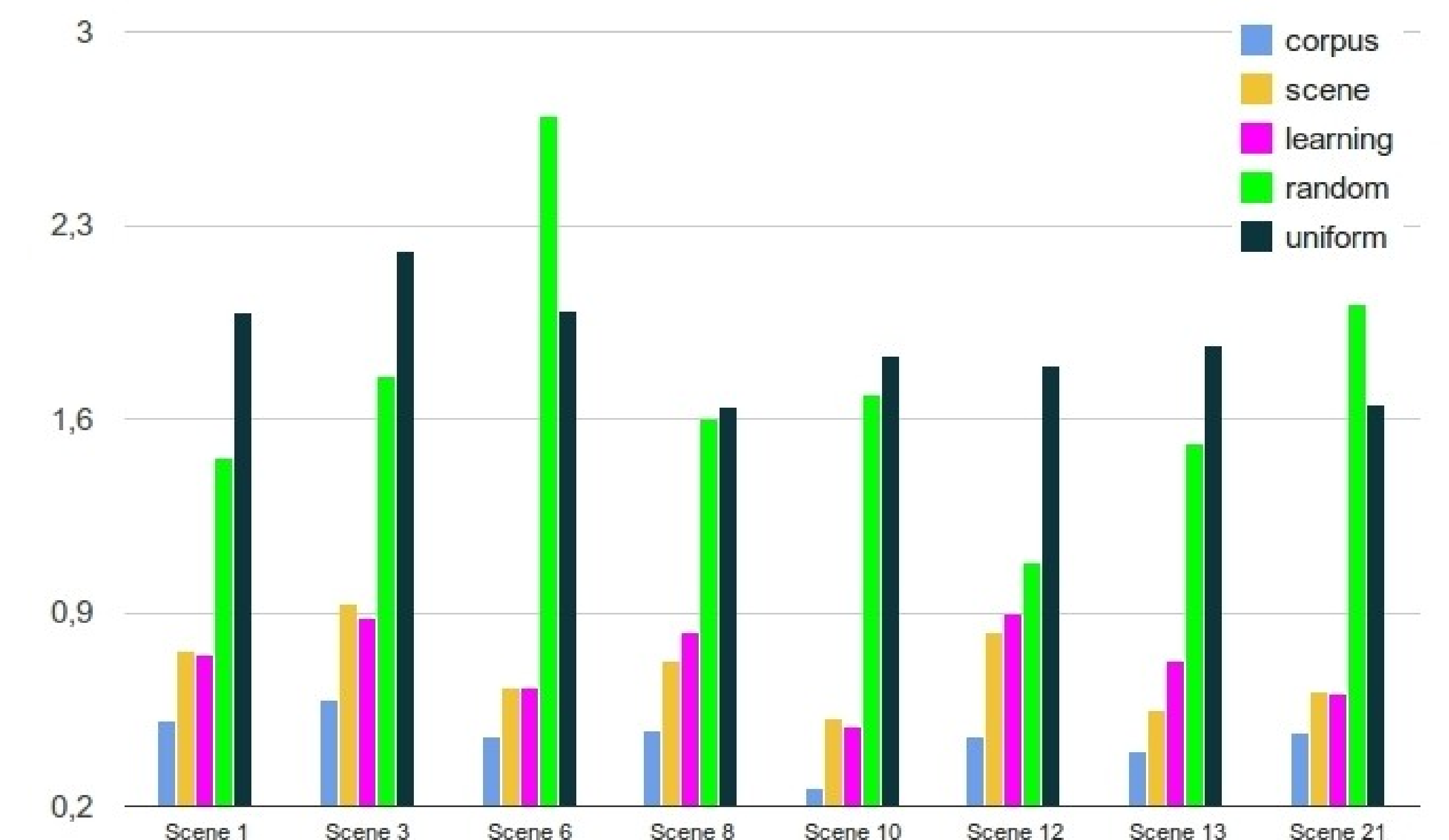
EVALUATION WITH PRECISION

Accuracy between the REs in the corpus and those generated using puse values computed from the scene, machine learned, random and uniform.

	Scene p_{use}	Learned p_{use}	Random p_{use}	Uniform p_{use}
Scene 1	85.75%	84.49%	17.95%	5.37%
Scene 3	82.81%	80.51%	9.89%	4.40%
Scene 6	90.11%	83.30%	4.13%	4.16%
Scene 8	86.52%	64.06%	16.32%	9.75%
Scene 10	89.49%	75.80%	7.56%	3.70%
Scene 12	80.21%	81.29%	57.09%	6.68%
Scene 13	89.98%	50.79%	9.30%	3.59%
Scene 21	92.13%	80.01%	8.45%	6.77%
Average	87.13%	75.03%	16.34%	5.55%

EVALUATION WITH CROSS ENTROPY

Cross-entropy between the corpus distribution and different runs of the algorithm.



- Our algorithm is able to generate different referring expressions for the same target with a frequency similar to that observed in corpora.
- Our results support the psycholinguistic theory that puts forward an egocentric explanation of language production (Keysar, 1998).