

### MOTIVATIONS

- Humans are social beings, and we learn language from interactions;
- A critical component of social interactions that language grounds to is the feedback provided by the caregivers;
- We study the role of corrective feedback in neural language learning through controlled computational experiments;
- Through ablation studies, our models can serve as proof of concept to verify mechanisms that are practically effective for machines, and generate hypotheses that are possible for cognitive learners.

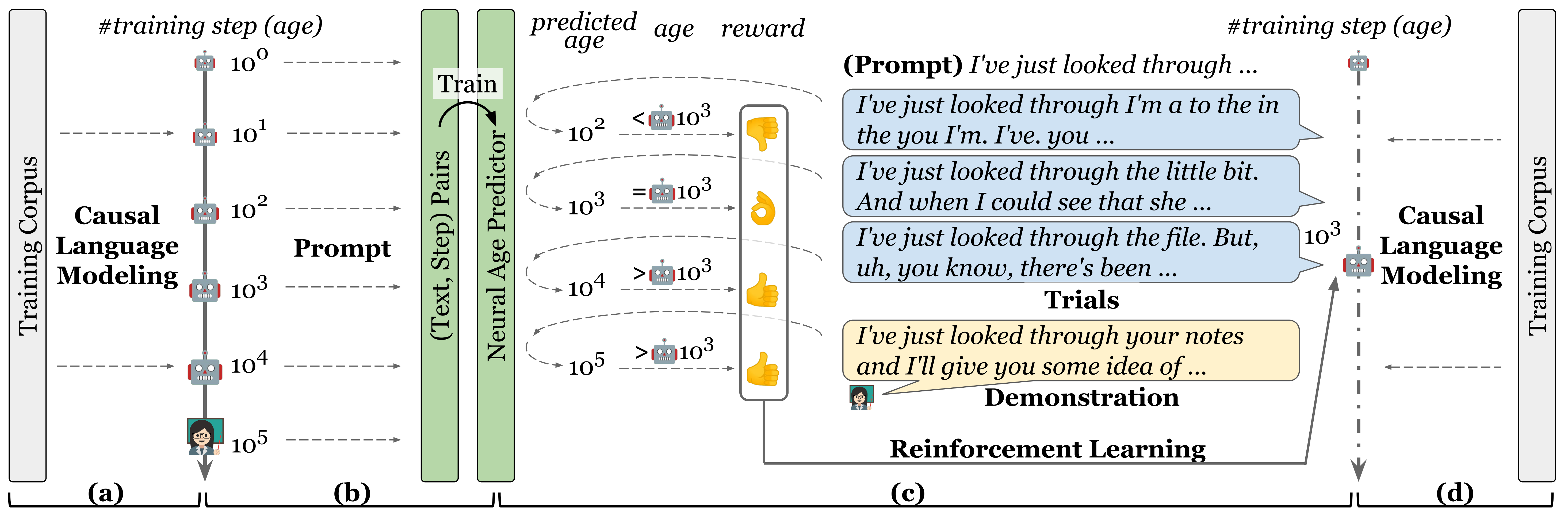
### KEY FINDINGS (TL;DR)

- Training language models from scratch with interactive language modeling by trials and demonstrations (TnD);
- TnD accelerates word acquisition for student models of equal and smaller numbers of parameters. Both trials and demonstrations matter;
- Teacher's choices of words influence students' word-specific learning efficiency;
- Students demonstrate a practice-makes-perfect effect, evident by a strong correlation between the word frequency trials and their respective learning curves.

### METHOD

- The trial and demonstration (TnD) learning framework with 3 components: student trials, teacher demonstrations, and an "age-conditioned" reward conditioned on language competence over time.
- Students do production-based learning: to produce an initial utterance, followed by the teacher model generating its version of the text as a demonstration.
  - We use a neural age predictor to estimate the expected training step  $\hat{n}$  when an utterance typically emerges, then normalized by the actual training step  $n$ .

### METHOD ILLUSTRATED: INTERACTIVE LANGUAGE LEARNING BY TRIALS AND DEMOS



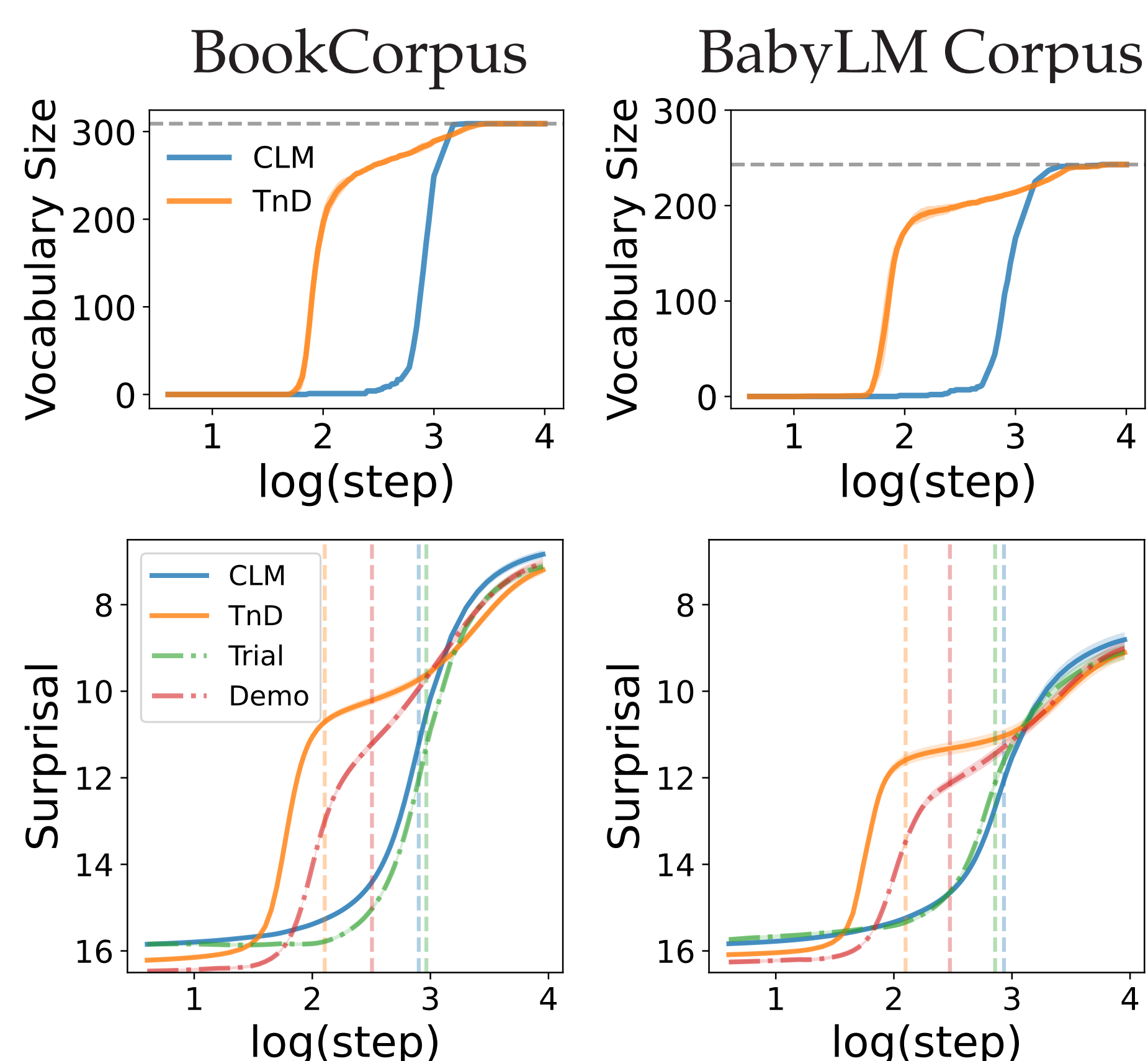
(a) Stage 1: Training a typical language model by causal language modeling.

(b) Stage 2: Training a neural age predictor from the trajectory of a typical language model.

(c) Stage 3: The student interactively learns from trials and demonstrations by a pre-trained teacher, score by an age-conditioned reward.

(d) Alternating between interactive learning and non-interactive learning.

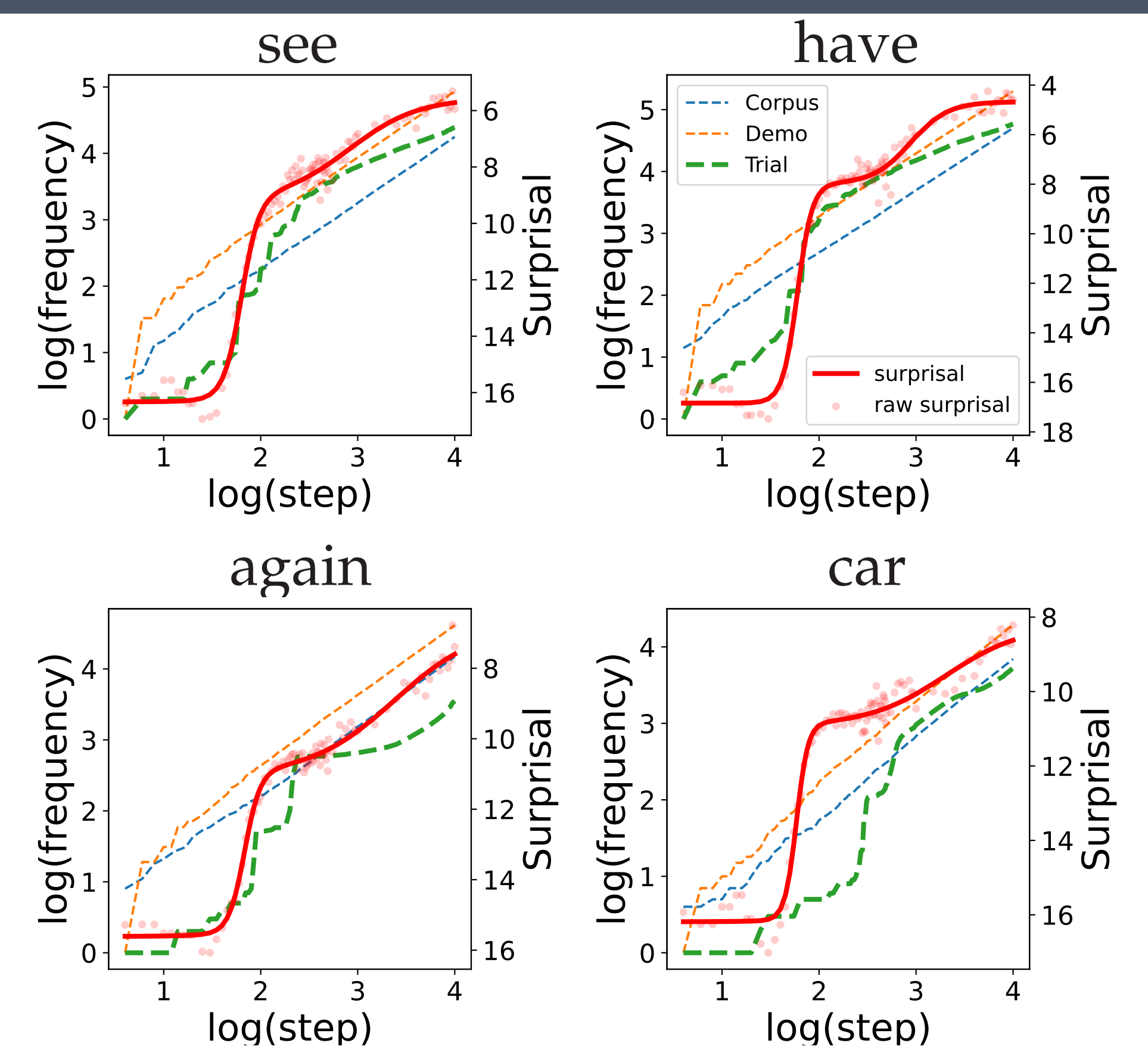
### MAIN RESULTS



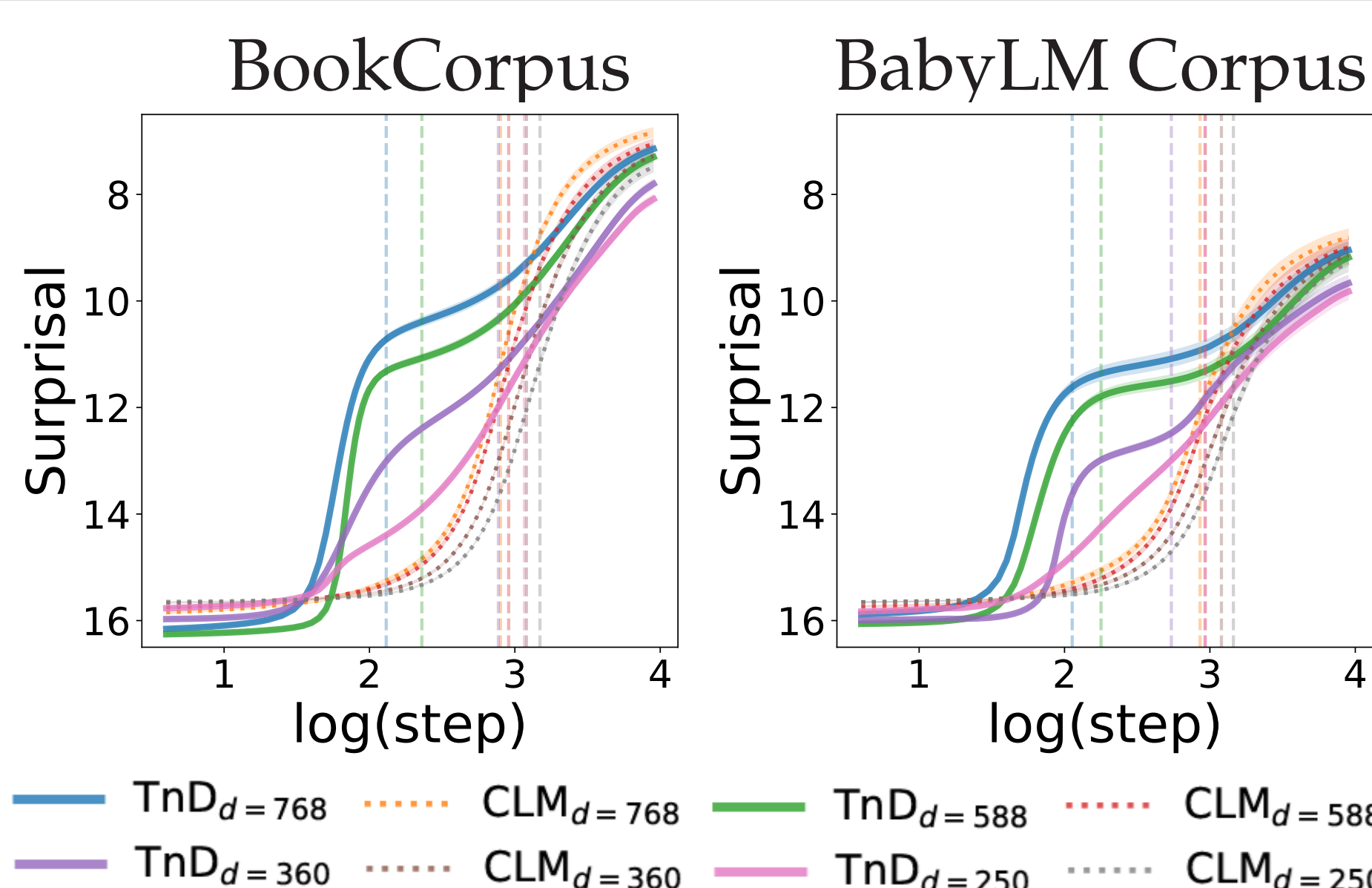
### METRICS

- Surprisal and learning curves. For each occurrence of a word  $w$ , the surprisal is  $-\log_2 P(w)$ , then averaged over all occurrences. We adopt a double-sigmoid function to fit the learning curve.
- Neural age of acquisition (nAoA@n%). The training step at which a surprisal cutoff of  $n\%$  between the minimum and maximum surprisal levels is achieved.
- Vocabulary size. The effective vocabulary size relative to a test set of vocabulary. A word is deemed acquired at step  $n$  if  $nAoA@0.50 \leq n$ .

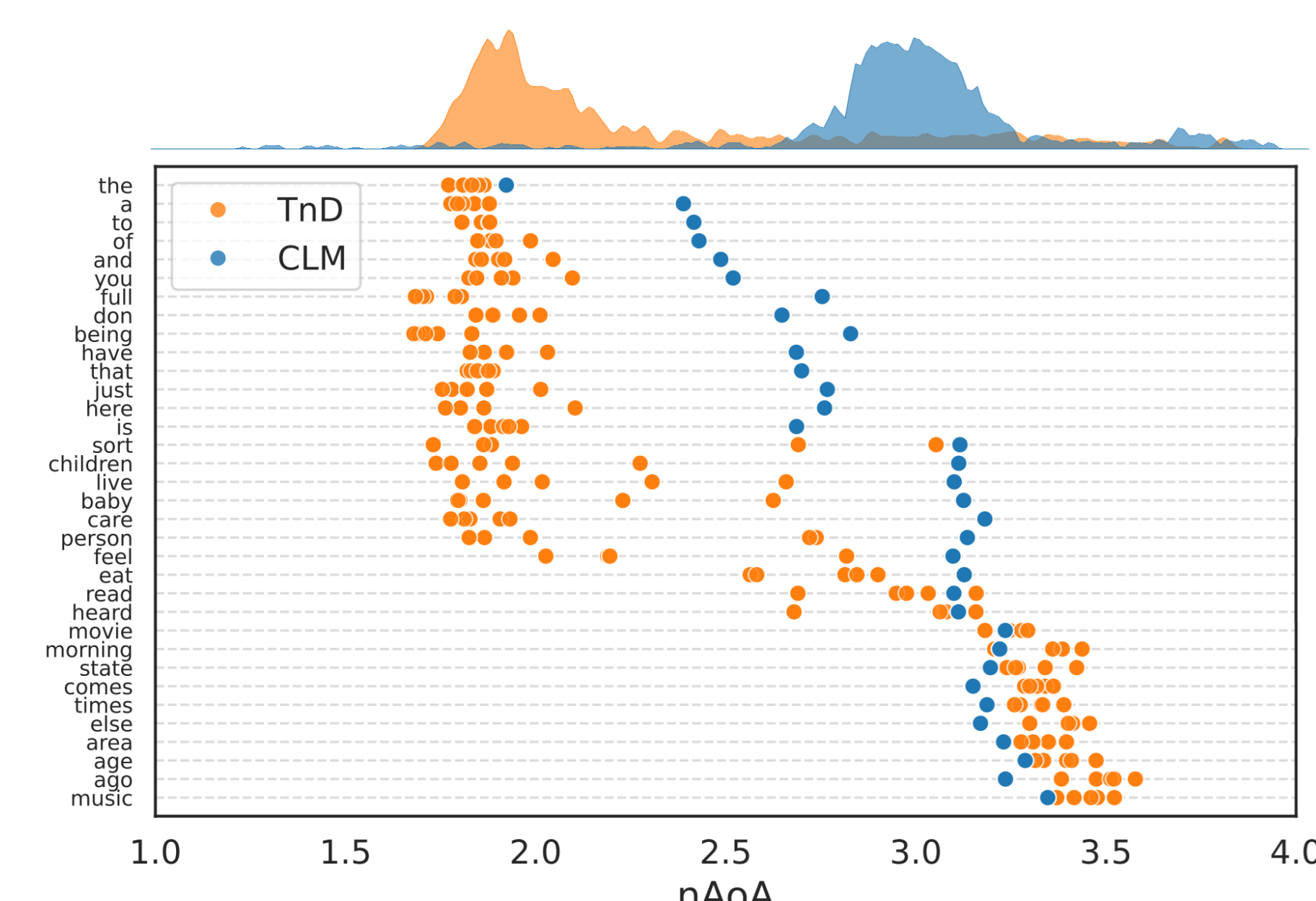
### STUDENT EFFECT



### DISTILLATION



### TRAJECTORY



### TEACHER EFFECT

