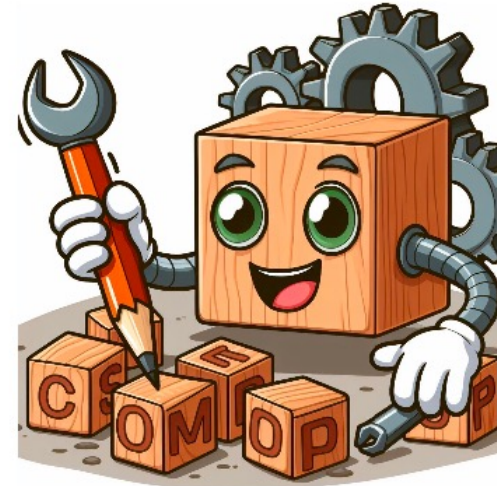


COMPO

Inductive Biases for Compositionality-capable Deep Learning Models of Natural Language

January 2024



Evolution of Large Language Models (LLMs)



- Outstanding successes on most NLP benchmarks have been substantially improved since 2018 and the birth of **transformers**

Position of the project



Natural language is compositional

- The amount of training data and the amount of computing resources required by current models is unrealistic

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Arnaud Rey
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LLMs demonstrate that it is possible to
produce grammatical sentences on the
basis of statistical learning and without
the slightest syntactic intervention

Are LLMs the end of the story?

Certainly not.

LLM training corpora are far too large
compared to the amount of information
a human is confronted with early in life
to learn to produce language

Statement 1: We are exposed to less linguistic information to develop our language skills

Statement 2: Our memory capacity is actually quite large

=> Which suggests that we may not need a very large learning set

Statement 3: With the development of our production capacity, we can **self-generate** linguistic forms, sometimes in a totally hidden way, within our inner language.

=> we can't rely on a simple assessment of what children are confronted with

- do you remember what you had for **dinner last night**?
 - do you remember **everything you've done** since you woke up this morning?
- => we are constantly memorizing a whole range of information **without** the slightest intention of memorizing it
- => all we have to do is **process** this information to create a transitory memory
- ⇒ most of these transitory memories disappear
- ⇒ However, if this information **repeats** itself (as is the case for regularities in our environment),
- these transitory memories are **reinforced**, becomes more and more **stable**,
and ultimately **structures** our subsequent mental states



The self-organizing consciousness

Pierre Perruchet and Annie Vinter

Université de Bourgogne, LEAD/CNRS, 21000 Dijon, France.

pierre.perruchet@u-bourgogne.fr annie.vinter@u-bourgogne.fr

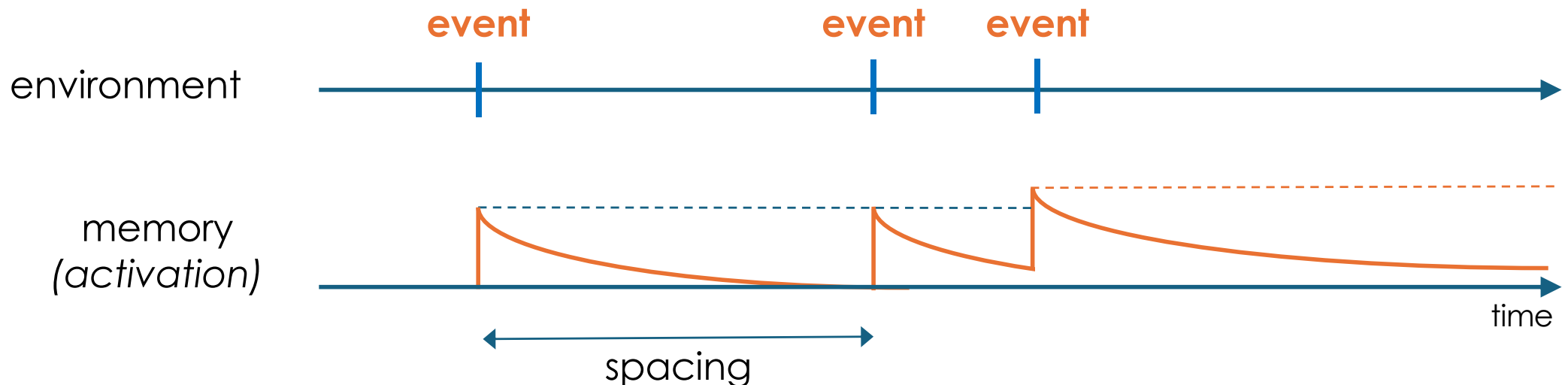
BEHAVIORAL AND BRAIN SCIENCES (2002) **25**, 297–388

- consciousness = attention = what we are **processing** at a given moment t
- we transiently memorize everything we **process**
(e.g., you remember almost everything you have processed/thought about
in the last few minutes/hours)
- this memorization is based on **elementary associative mechanisms**
(that we don't yet fully understand)
- these associative mechanisms enable us to **extract regularities** from the environment
whose main property is to be **repeated** and thus be increasingly memorized

Question : Can we estimate the **time** during which transient memory traces **survive**?

=> for an environmental regularity to be **memorized**, it is necessary that:

1. the transient memory of its first occurrence must **survive** until its **first repetition**
2. the **interval between two repetitions** must not be too great, so that the memory trace benefits from the repetitions to consolidate.



- starting point: **Hebb's repetition task** (1961)

=> serial recall of sequences of digits

=> a sequence is repeated every 3 sequences

- Hebb **naming** task (Rey et al., 2020 - QJEP)

=> read aloud isolated letters

=> a triplet of **letters** is repeated regularly

=> measures RT evolution over repetitions

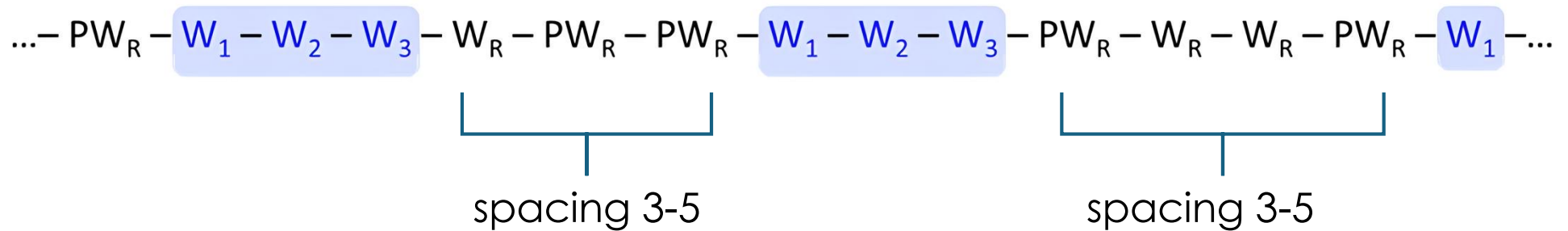
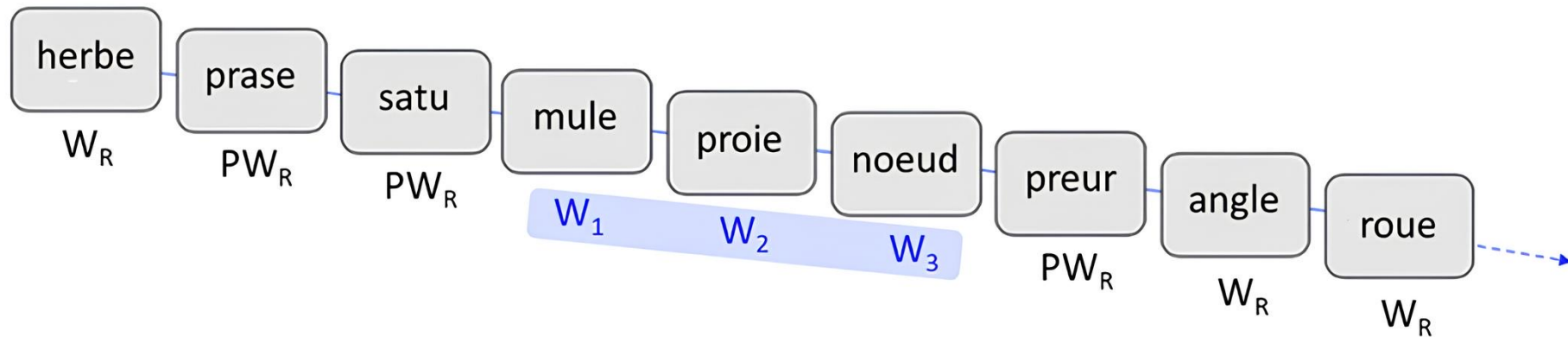
- Hebb **lexical decision** task (Pinto Arata, ..., Rey, 2024 - QJEP)

=> lexical decision task on items presented one by one

=> a triplet of three items is repeated

=> repetition **spacing**: between 3 and 5 item fillers





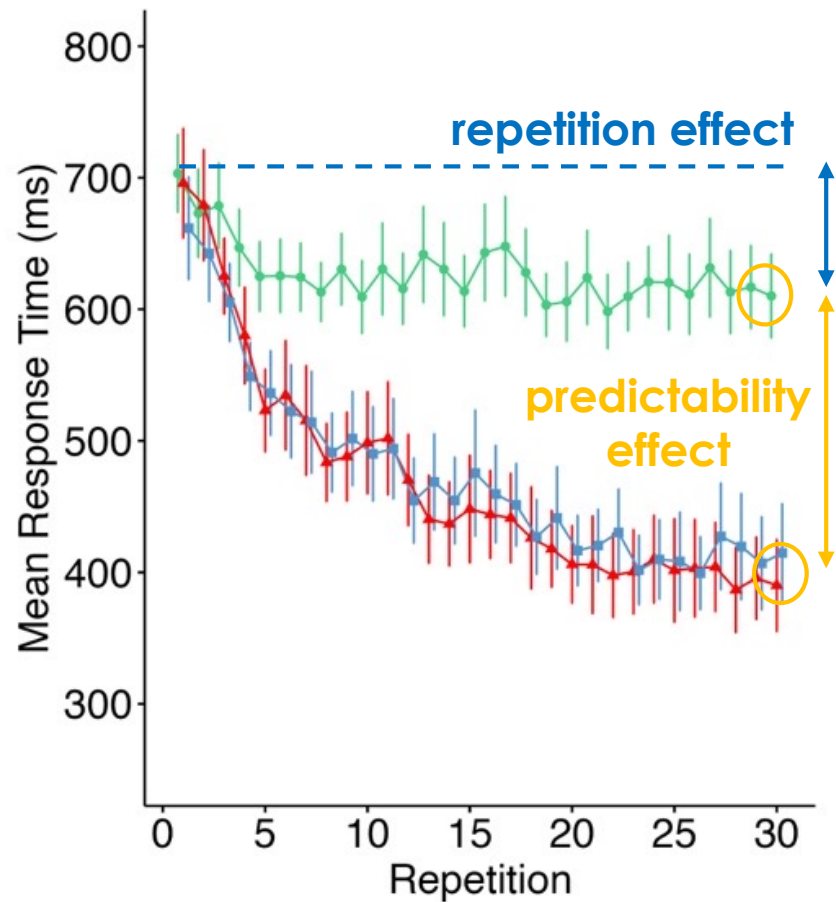
Present study (currently submitted):

- Hebb lexical decision task (Pinto Arata, ..., Rey, 2024 - QJEP)
- 5 spacing conditions :

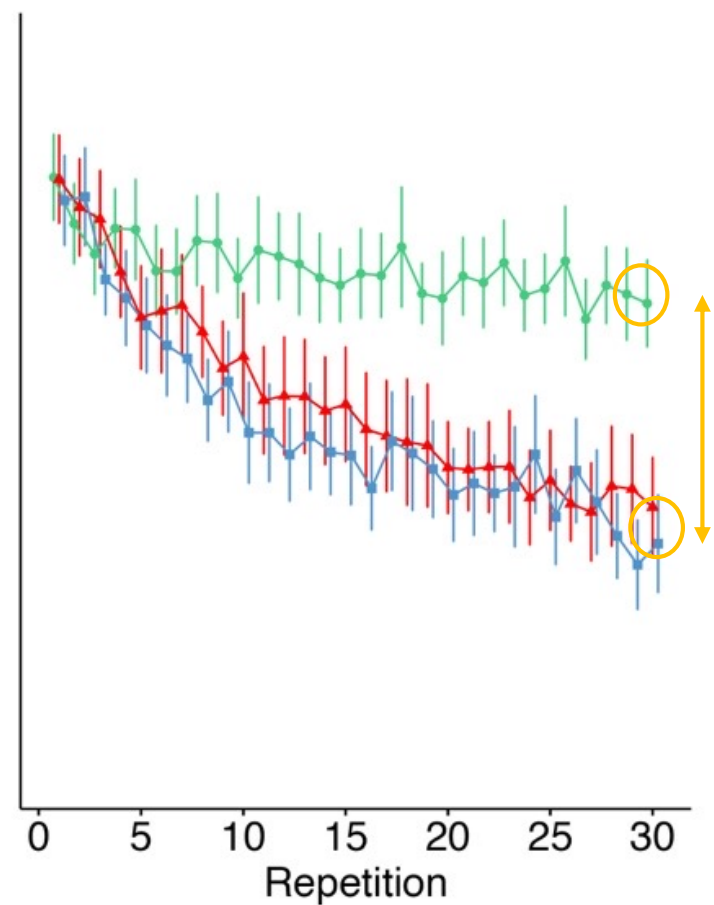
| | |
|----------|-------|
| => 3-5 | = F4 |
| => 6-8 | = F7 |
| => 9-11 | = F10 |
| => 19-21 | = F20 |
| => 29-31 | = F30 |
| => 59-61 | = F60 |
- 420 participants, 84 per condition
- online experience: www.prolific.co

Position ● 1 ▲ 2 ■ 3

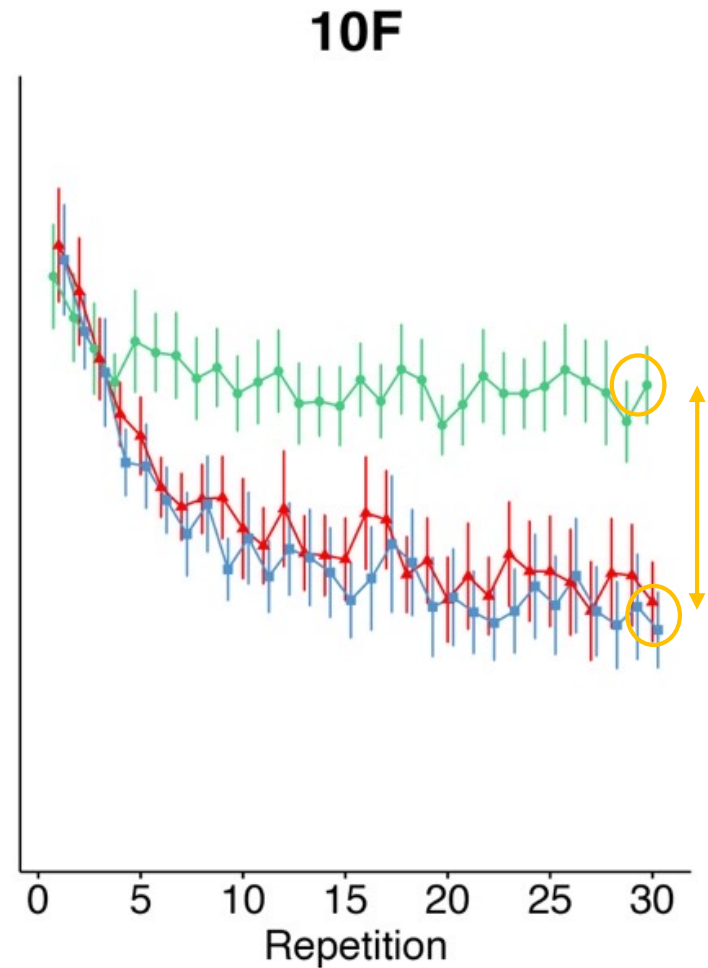
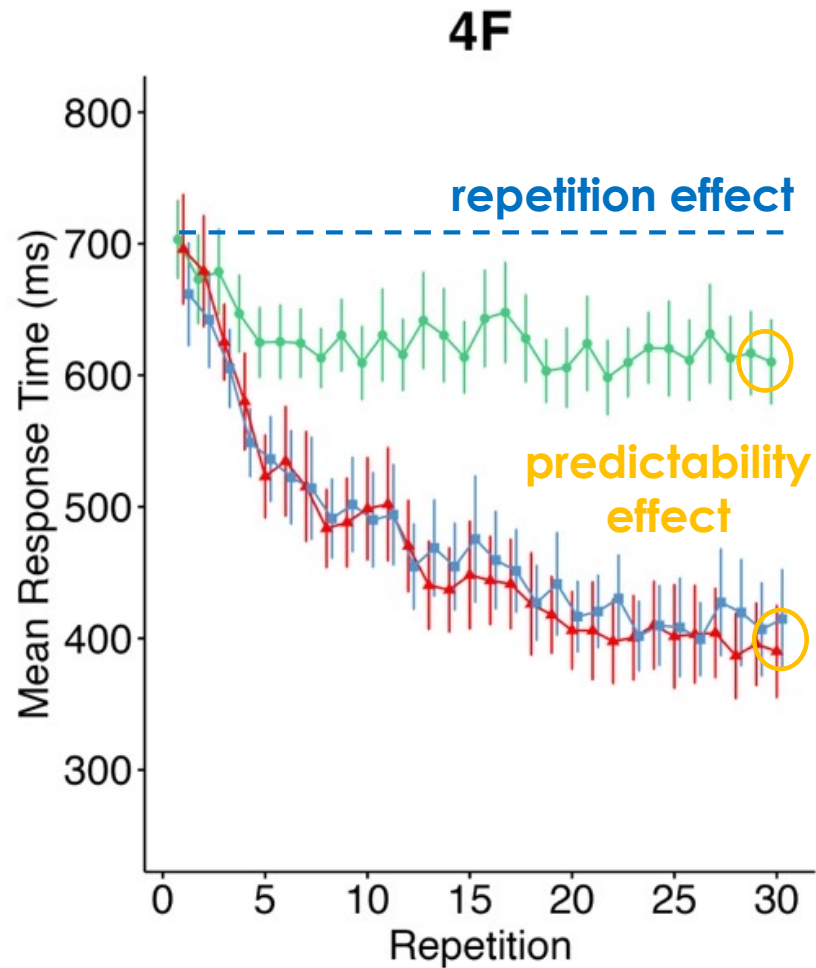
4F

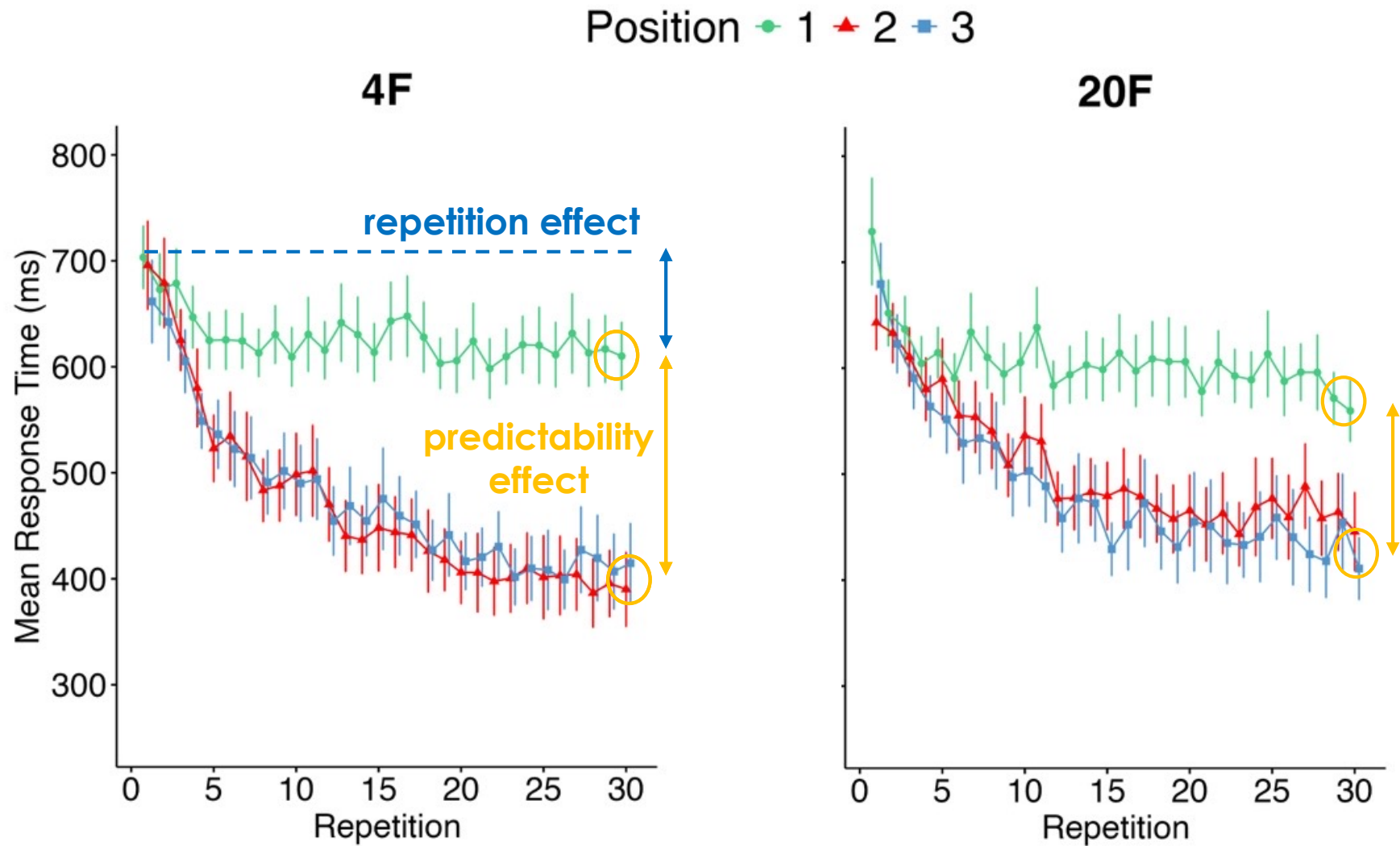


7F

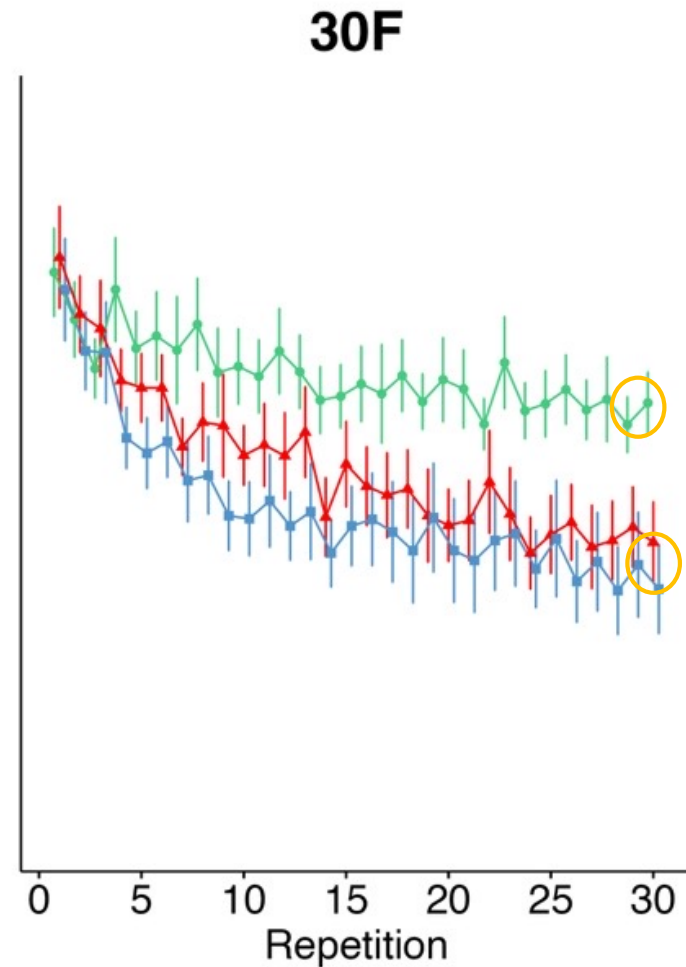
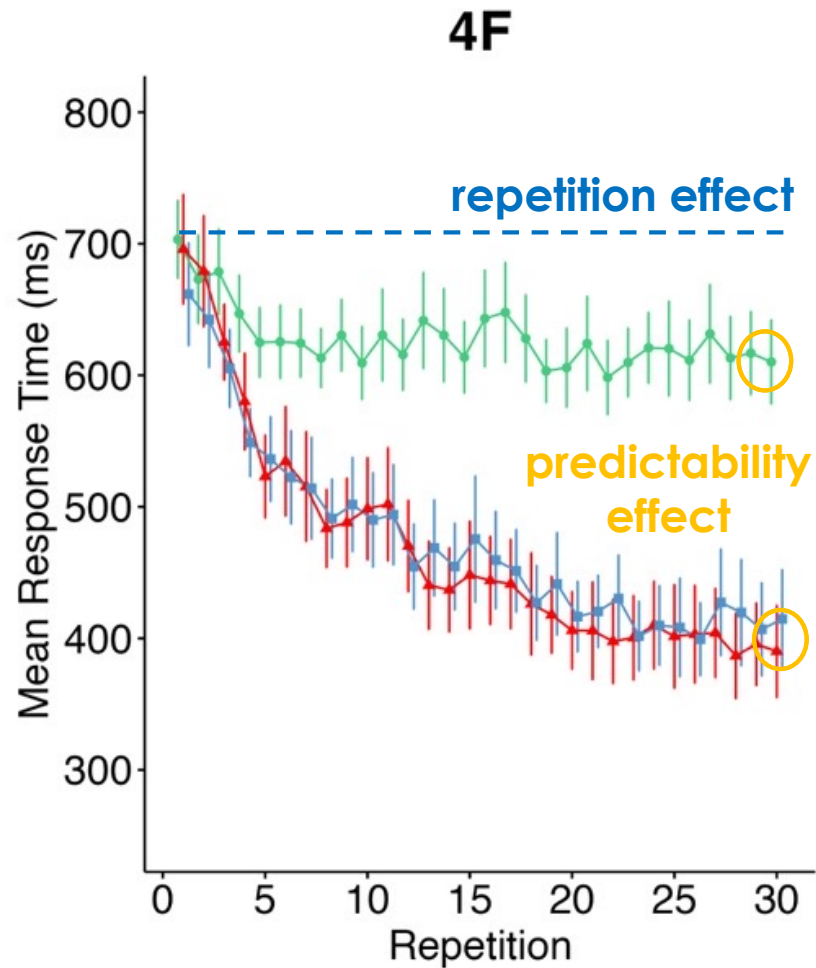


Position ● 1 ▲ 2 ■ 3



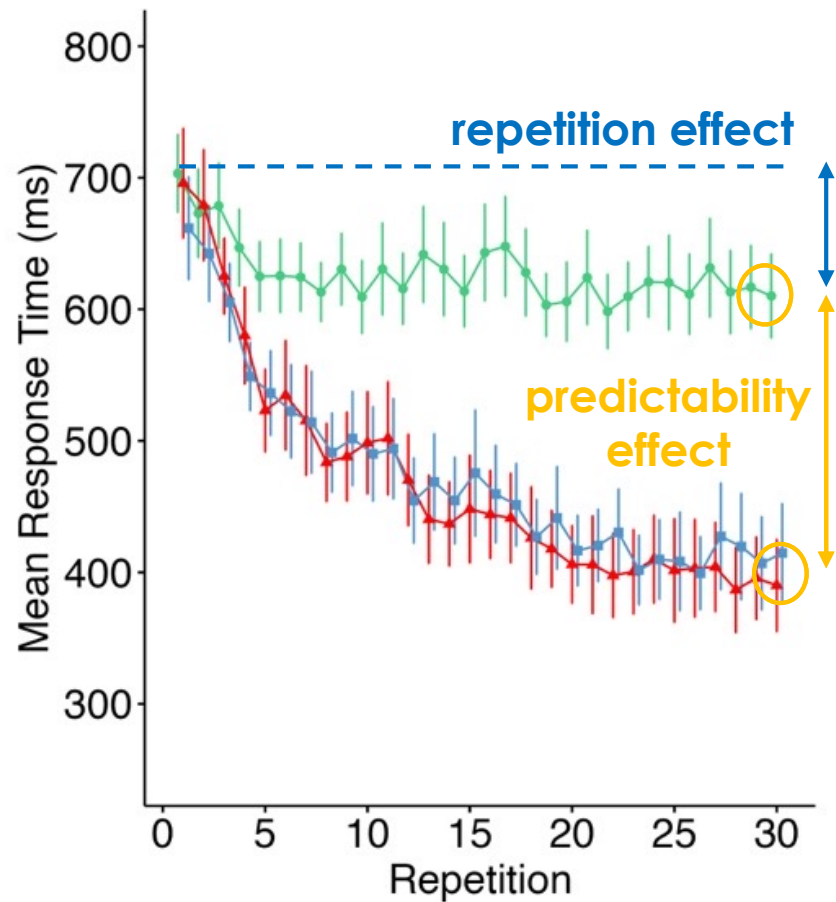


Position ● 1 ▲ 2 ■ 3



Position ● 1 ▲ 2 ■ 3

4F



60F

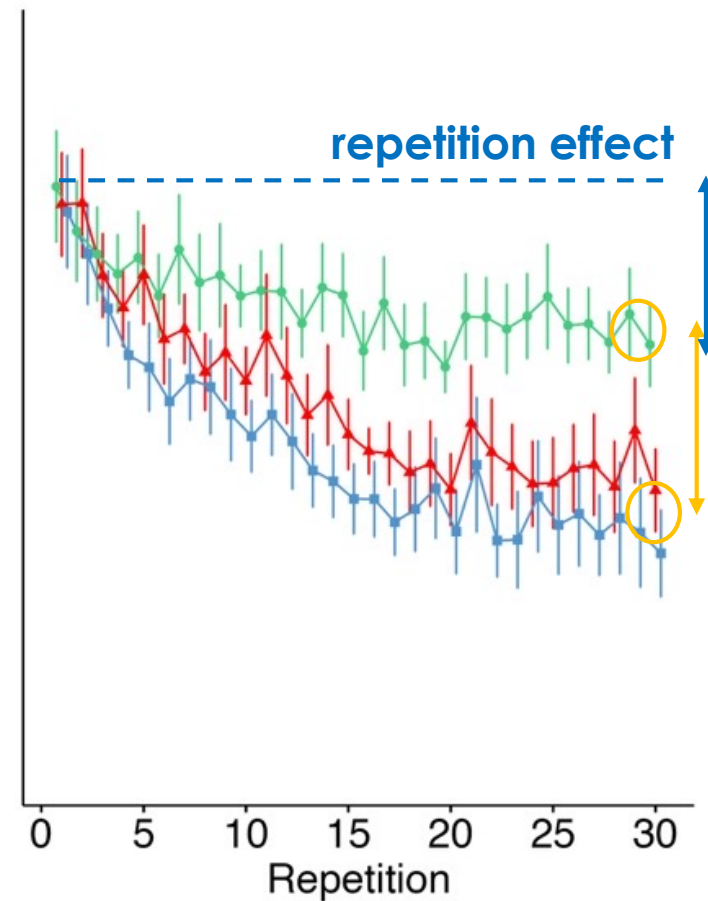
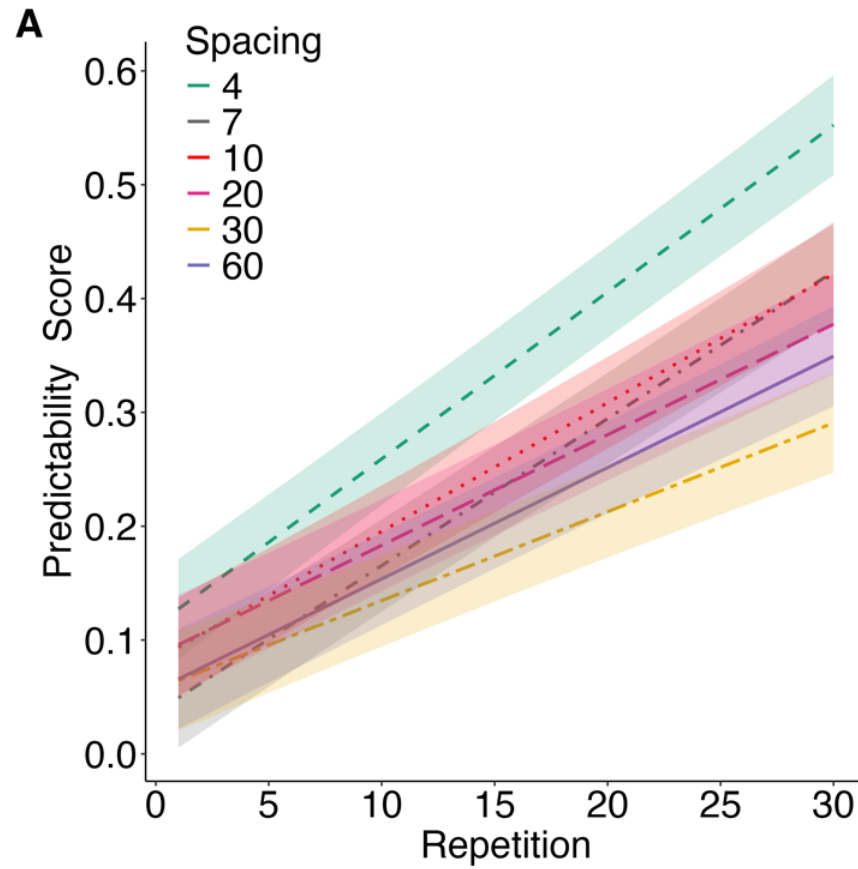


Figure 1 displays six line graphs showing Mean Response Time (ms) versus Repetition (0 to 30) for different frequencies (4F, 7F, 10F, 20F, 30F, 60F). Each graph contains three data series: green, red, and blue, all showing a decrease in response time as repetition increases. The y-axis ranges from 300 to 800 ms, and the x-axis ranges from 0 to 30 repetitions. Error bars are present for each data point.

| Frequency | Series | Approx. Mean Response Time (ms) at Repetition 0 | Approx. Mean Response Time (ms) at Repetition 30 |
|-----------|--------|---|--|
| 4F | Green | ~700 | ~600 |
| | Red | ~700 | ~400 |
| | Blue | ~680 | ~420 |
| 7F | Green | ~700 | ~600 |
| | Red | ~700 | ~450 |
| | Blue | ~680 | ~420 |
| 10F | Green | ~700 | ~580 |
| | Red | ~700 | ~420 |
| | Blue | ~680 | ~400 |
| 20F | Green | ~730 | ~550 |
| | Red | ~650 | ~450 |
| | Blue | ~680 | ~420 |
| 30F | Green | ~680 | ~580 |
| | Red | ~680 | ~480 |
| | Blue | ~650 | ~450 |
| 60F | Green | ~680 | ~580 |
| | Red | ~680 | ~480 |
| | Blue | ~650 | ~420 |

Predictability score : $\log RT_{2/3} - \log RT_1$



=> power law

Statement 2: Our memory capacity is actually quite large

=> Which suggests that we may not need a very large learning set

Statement 3: With the development of our production capacity, we can **self-generate** linguistic forms, sometimes in a totally hidden way, within our inner language.

=> self-generation produces repetitions = reinforced memory traces

=> the poverty of the stimulus is compensated by the richness of self-productions

Statement 4: This is not the end of the story.

We need to develop the fourth stage of connectionist modeling

Topics in Cognitive Psychology, 2024, 124, Juin 2024 / June 2024

Associations are all we need

Arnaud Rey *

CNRS & Aix-Marseille Université, Marseille, France

ABSTRACT

In this opinion article, I argue that the time has come for a unified radical associationism that is built around the assumption that associations are all we have and probably all we need to account for mental activities. This radical associationism should be able to *merge* the fields of associative, statistical and Hebbian learning and unify these theoretical and empirical approaches. A direct consequence of adopting such a theoretical stance is a revision of several key psychological concepts (e.g., the notion of attention) based on neurobiological ones, leading to a unified neuropsychological theorization of mental activities.

Keywords: Associationism, statistical learning, Hebbian learning

2 Connectionist Models of Cognition

MICHAEL S. C. THOMAS AND JAMES L. MCCLELLAND

First age

Second age

Third age

