

EVENT SCHEMAS: LEARNING AND USE IN HUMANS AND RECURRENT NETWORKS

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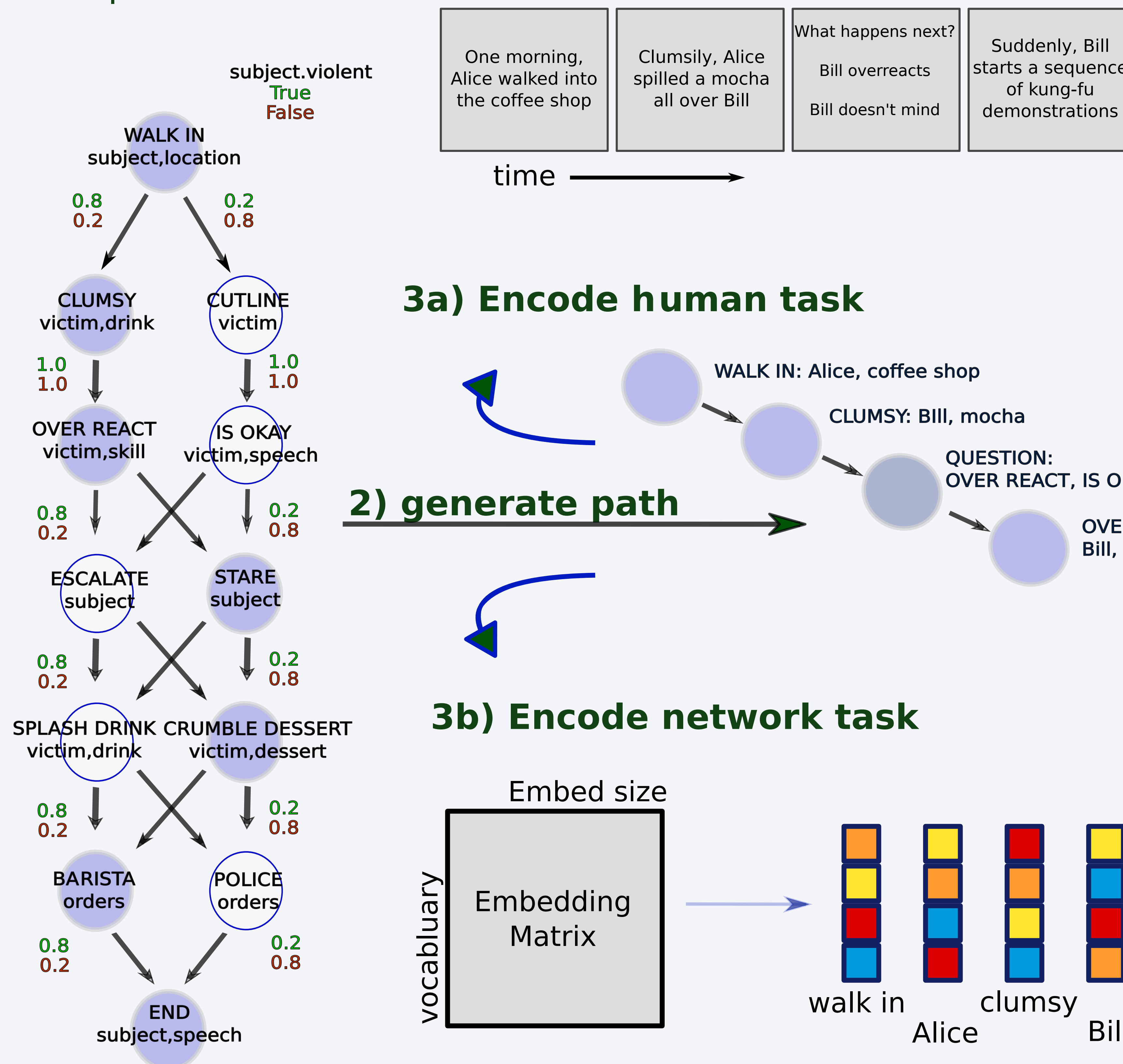
Schemas

- * are the scaffolding of memory
- * constructed from multiple episodes
- * afford generalization
- * support encoding
- * how are they learned and used?

Approach:

- * algorithmically generate narratives with:
 - long range probabilistic dependencies
 - filler dependent transition

1) sample fillers: fix probabilities

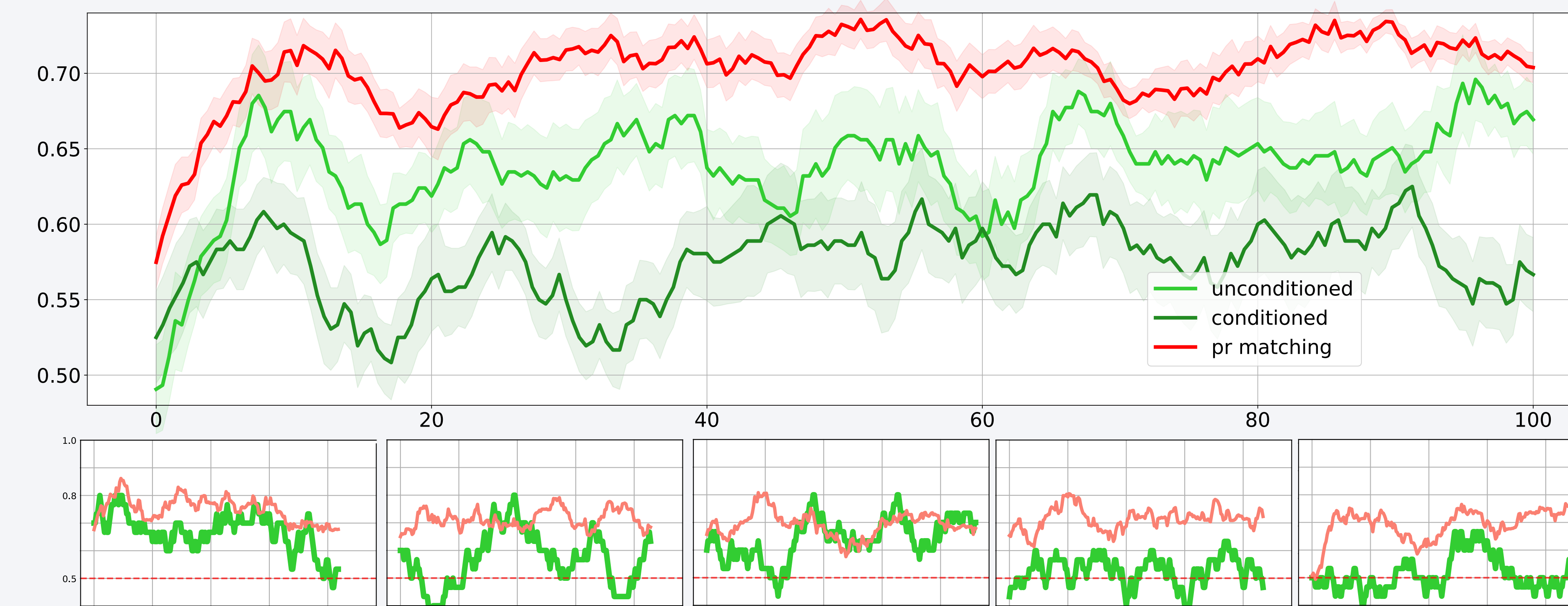


Humans

stop and ask 2AFC. what happens next?
all experiments on mturk. n=25 above catch question thresh. chance is 50%

is learning possible in our task?

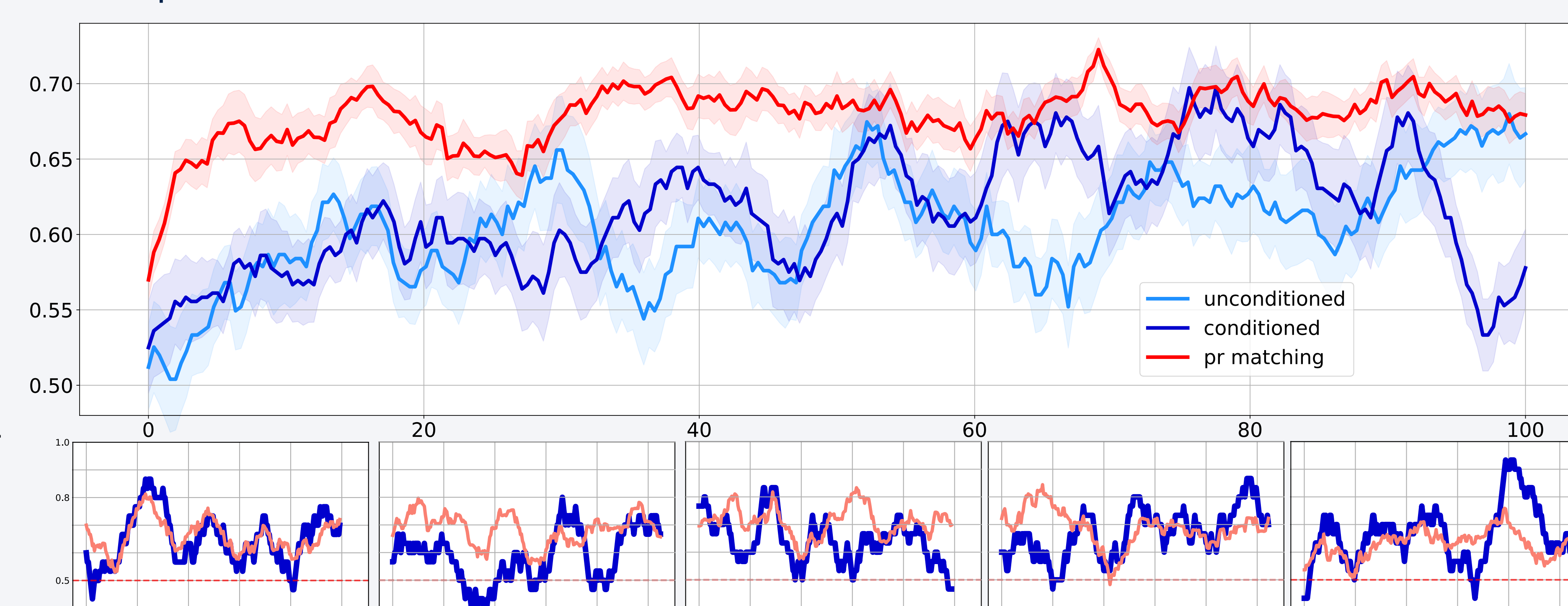
only subject and victim names changing



subjects approached probability matching performance on unconditioned transitions more so than on conditioned transitions.

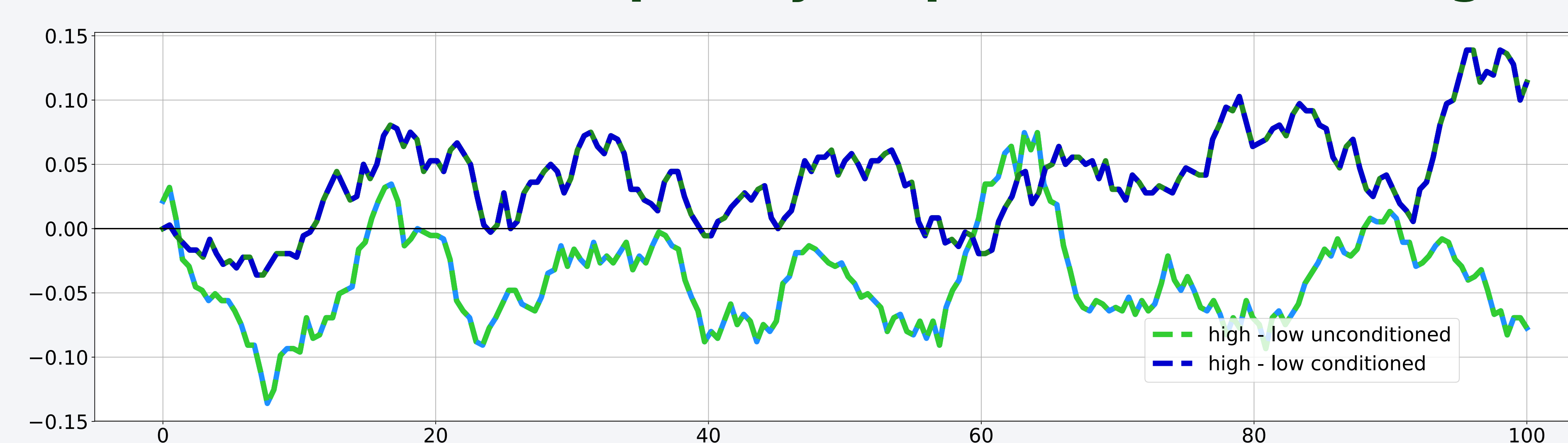
high filler complexity

4096 possible different combinations of fillers



learning is still possible in high filler complexity

does surface complexity help or hinder learning?

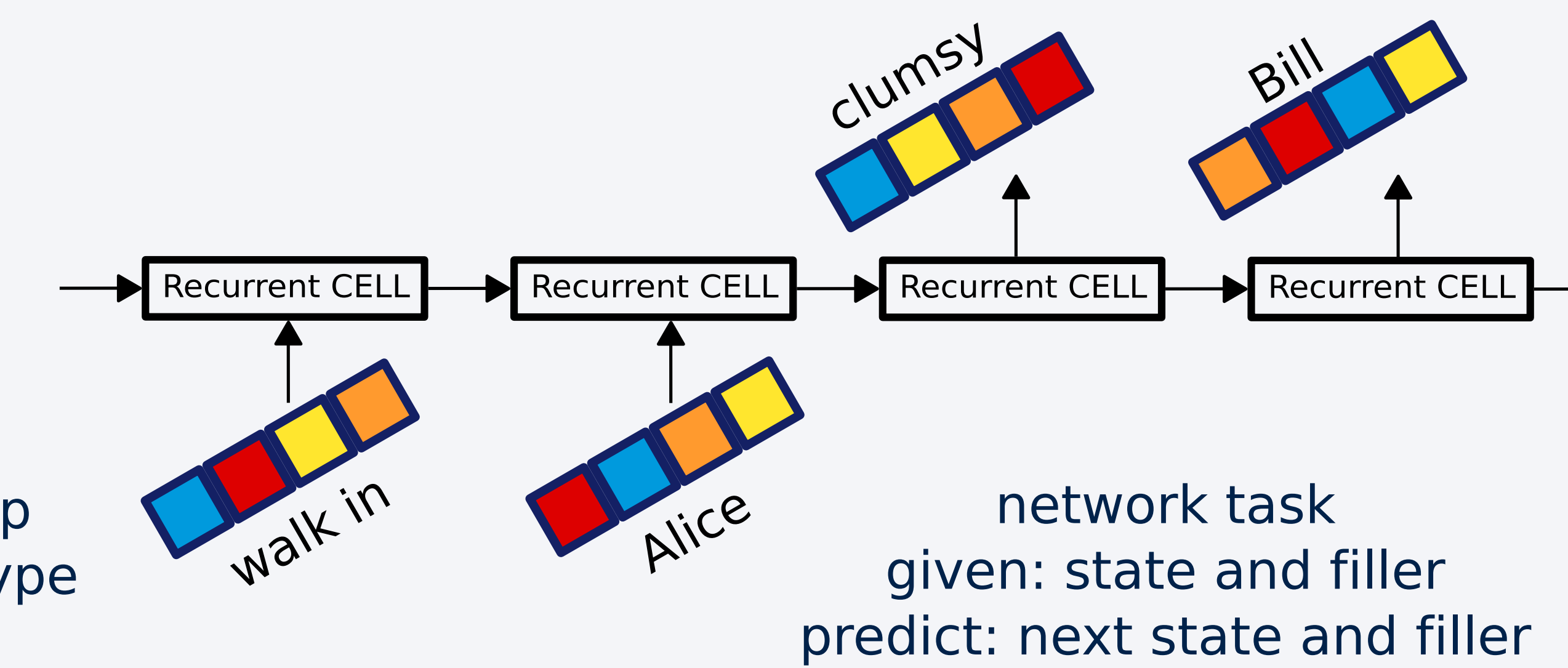


surface complexity helps on conditioned transitions but hinders on unconditioned transitions

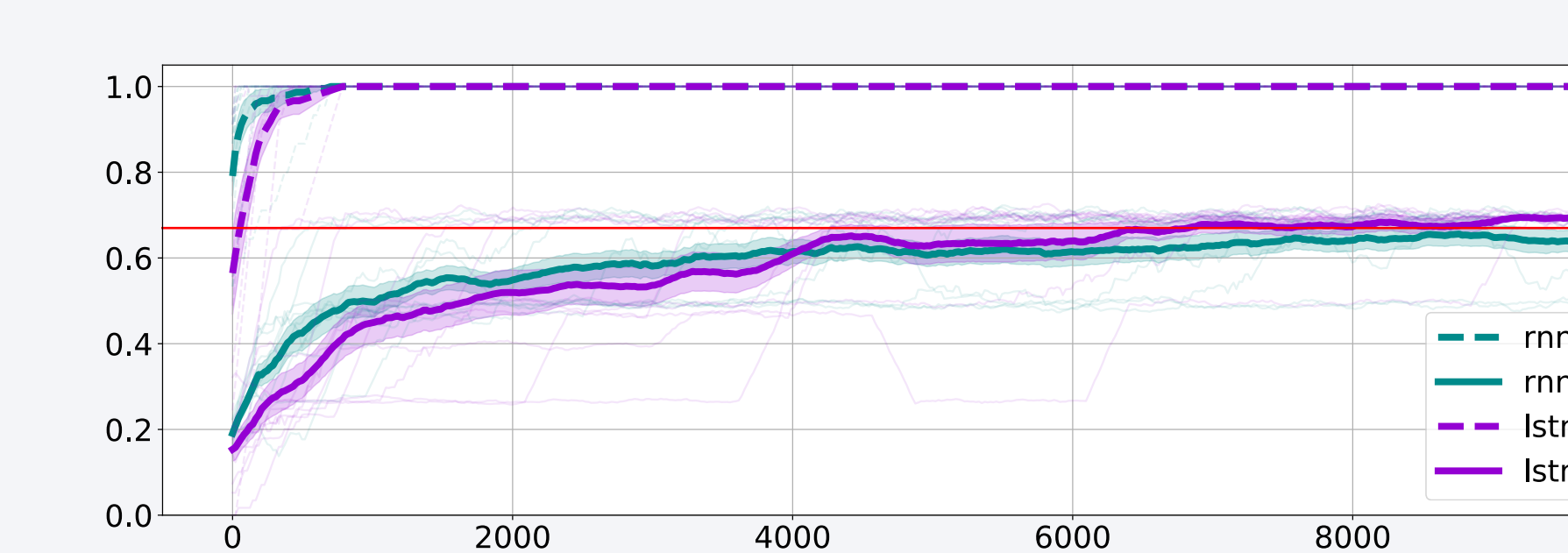
Networks

RNN vs LSTM

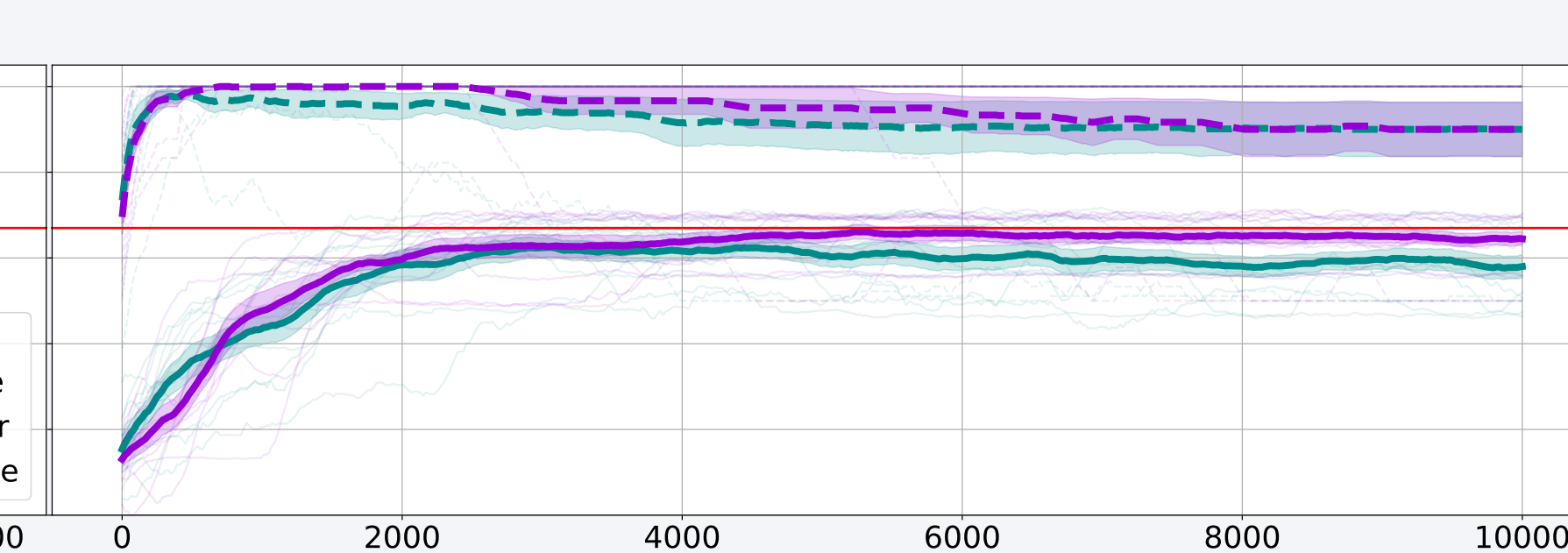
all experiments on laptop
n=10 of each network type
chance is 0.05%



Unconditioned



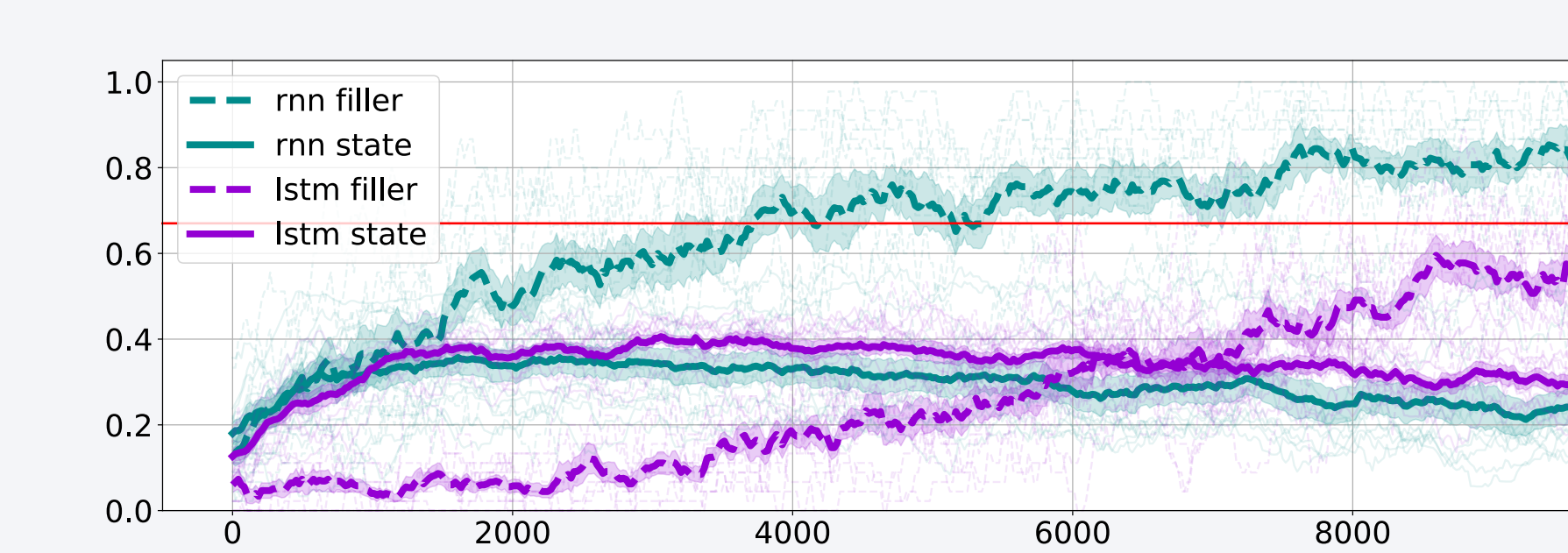
Conditioned



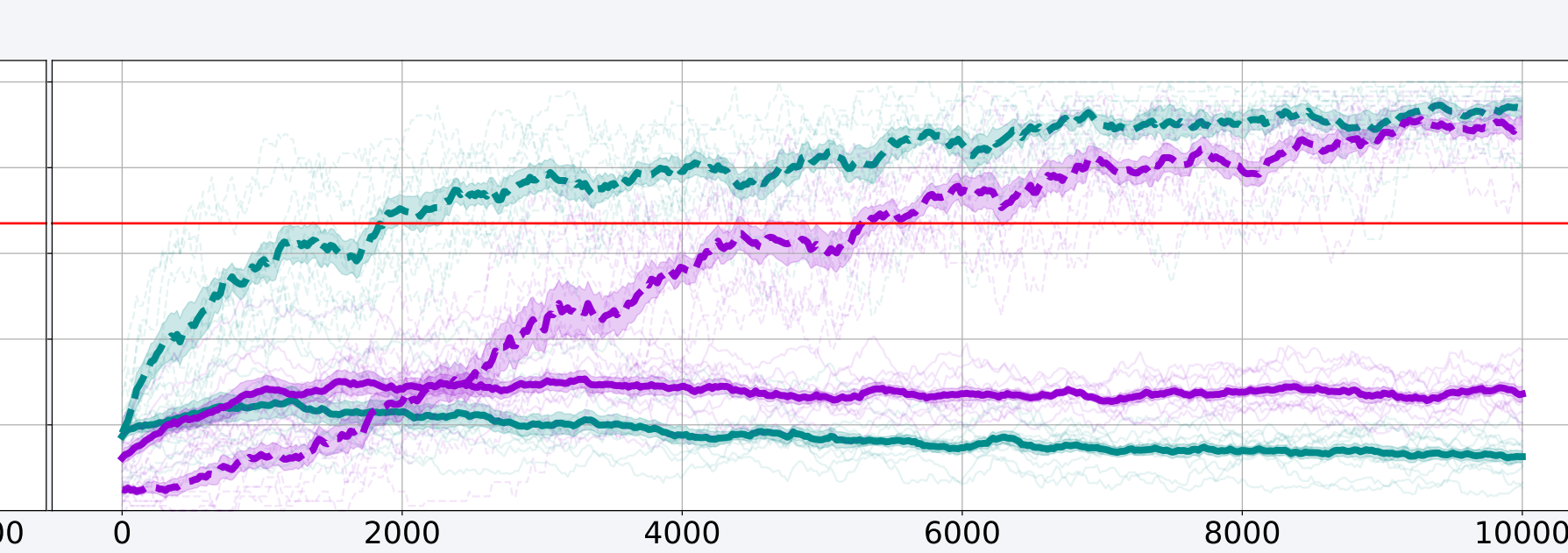
network performance mirrors human performance. both networks probability matched on next state questions and scored close to perfect on filler questions

Future Directions

Unconditioned



Conditioned



- * generalization task: filler vectors randomized.
- * randomizing inhibits probability matching behavior.
- * networks at chance in conditioned transitions where filler information is required for next state questions.
- * began investigating the impact of different learning regimes: blocked versus interleaved learning, and curriculum learning. how do these influence learning dynamics, task solutions and latent representations?

Take home

- * validation of new task for studying schemas
- * naturalistic complexity helps learning
- * different mnemonic architectures have different learning dynamics and task solutions