

Neural networks

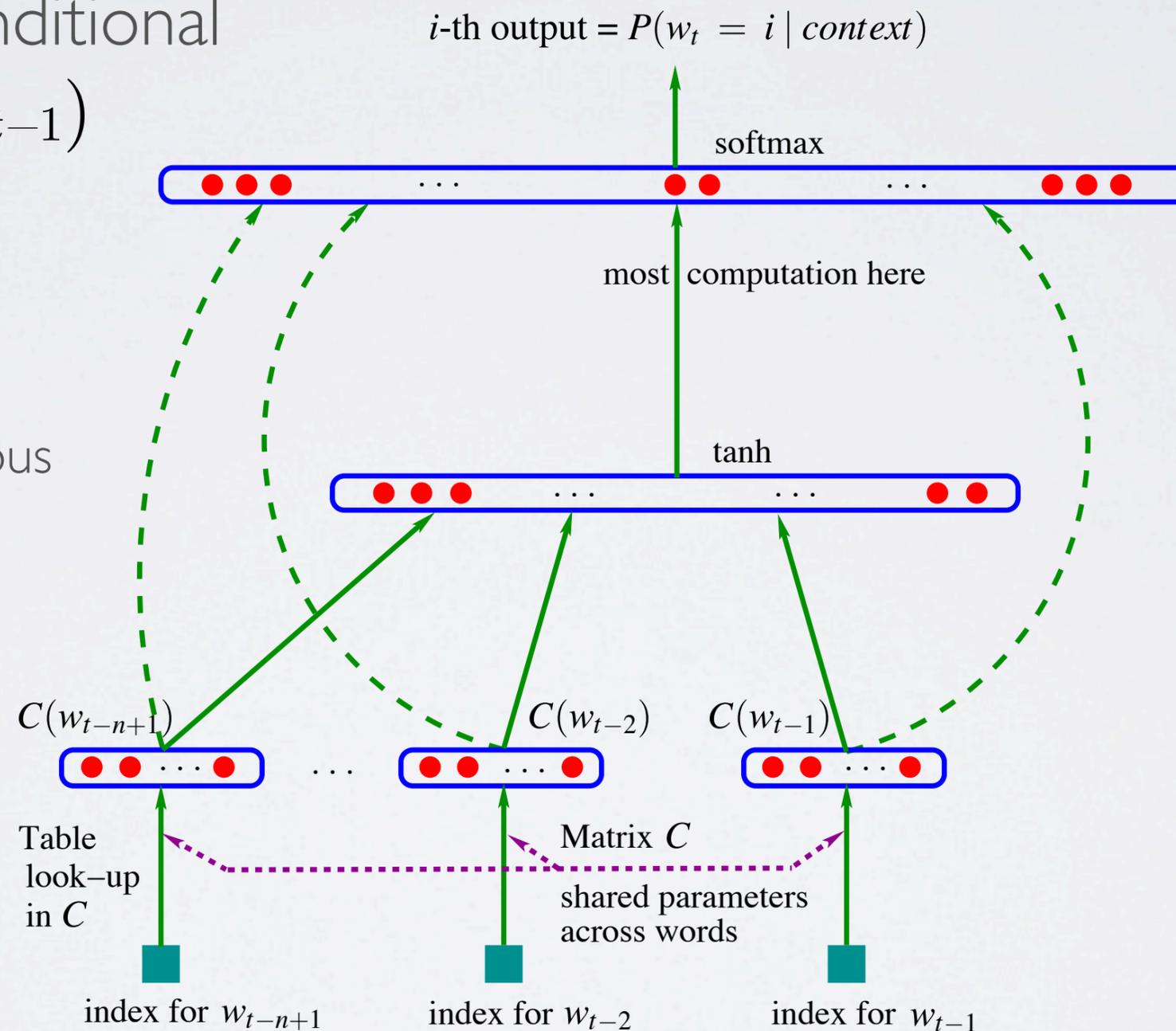
Natural language processing - hierarchical output layer

NEURAL NETWORK LANGUAGE MODEL

Topics: neural network language model

- Solution: model the conditional $p(w_t \mid w_{t-(n-1)}, \dots, w_{t-1})$ with a neural network

- ▶ learn word representations to allow transfer to n -grams not observed in training corpus



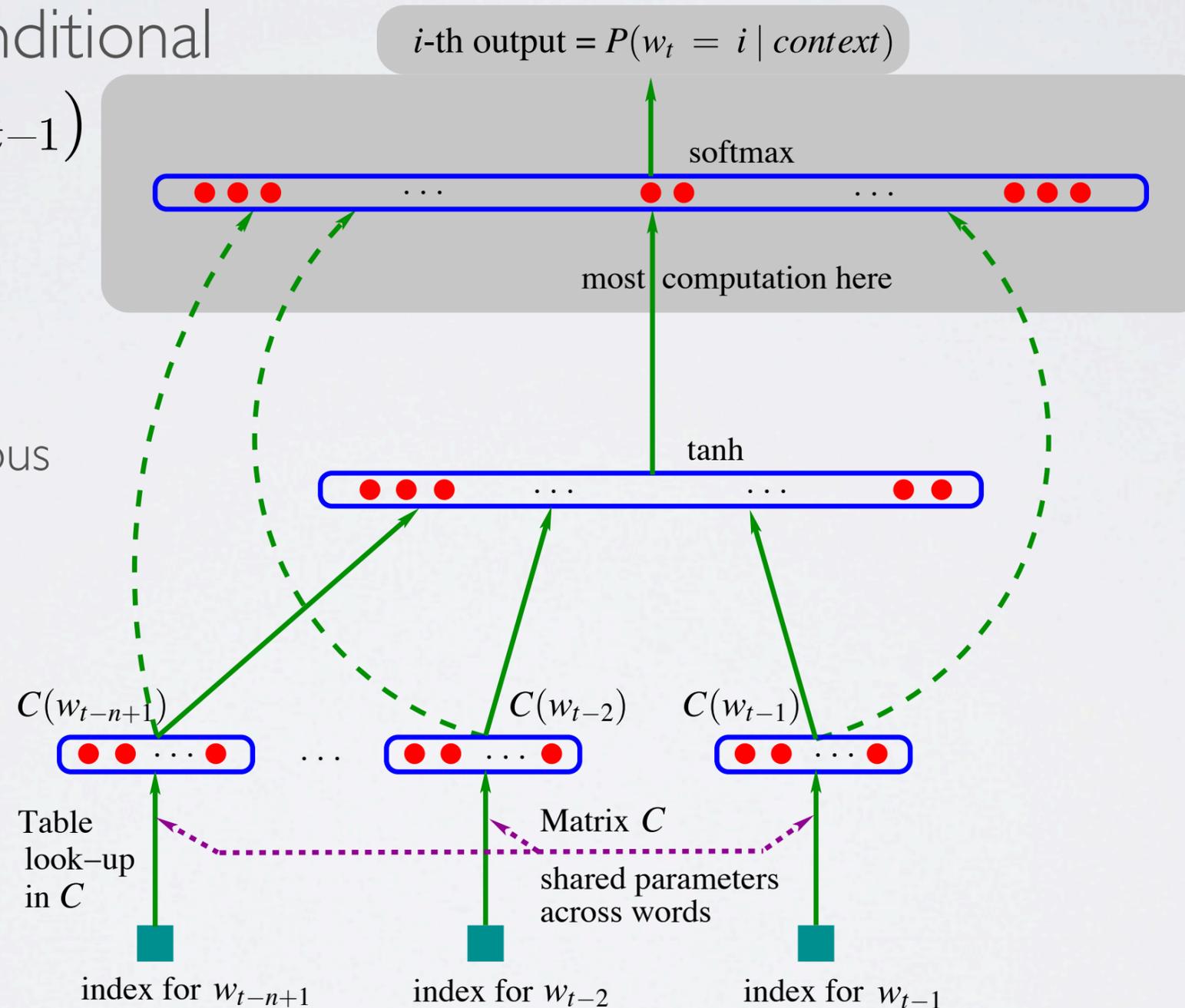
Bengio, Ducharme,
Vincent and Jauvin, 2003

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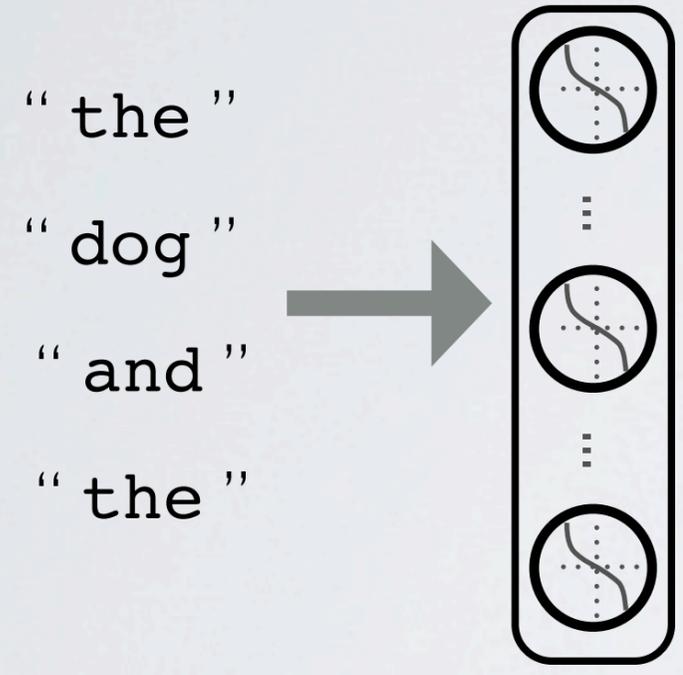


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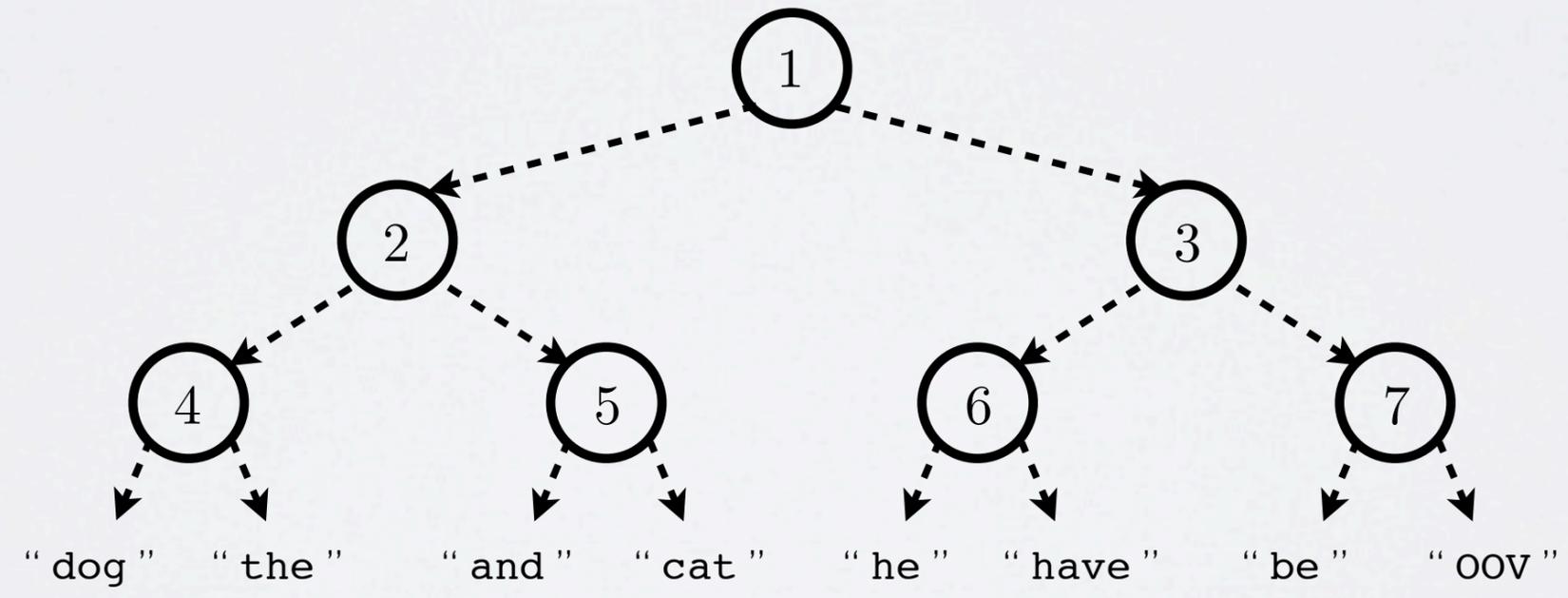
NEURAL NETWORK LANGUAGE MODEL

Topics: hierarchical output layer

- Example: [“ the ”, “ dog ”, “ and ”, “ the ”, “ cat ”]



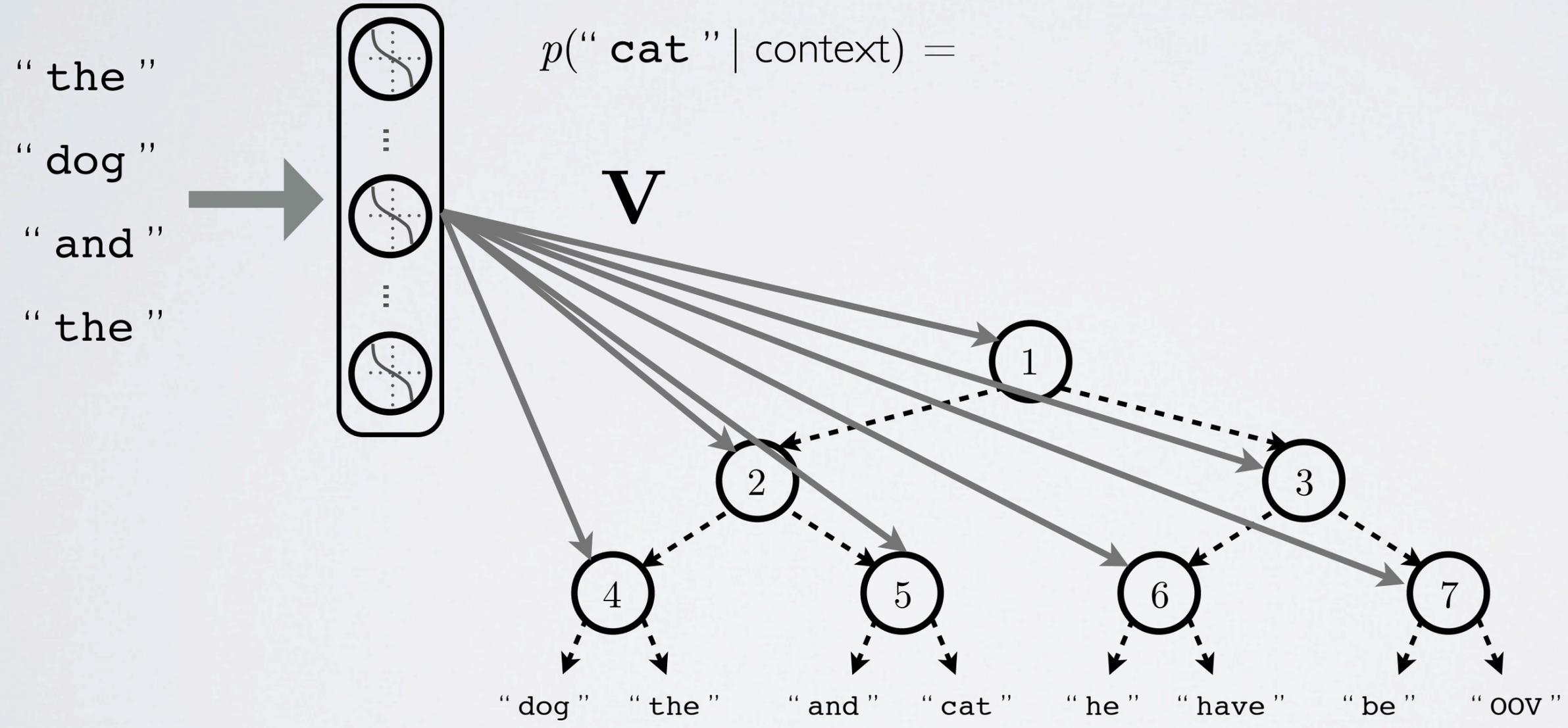
$$p(\text{“ cat ”} \mid \text{context}) =$$



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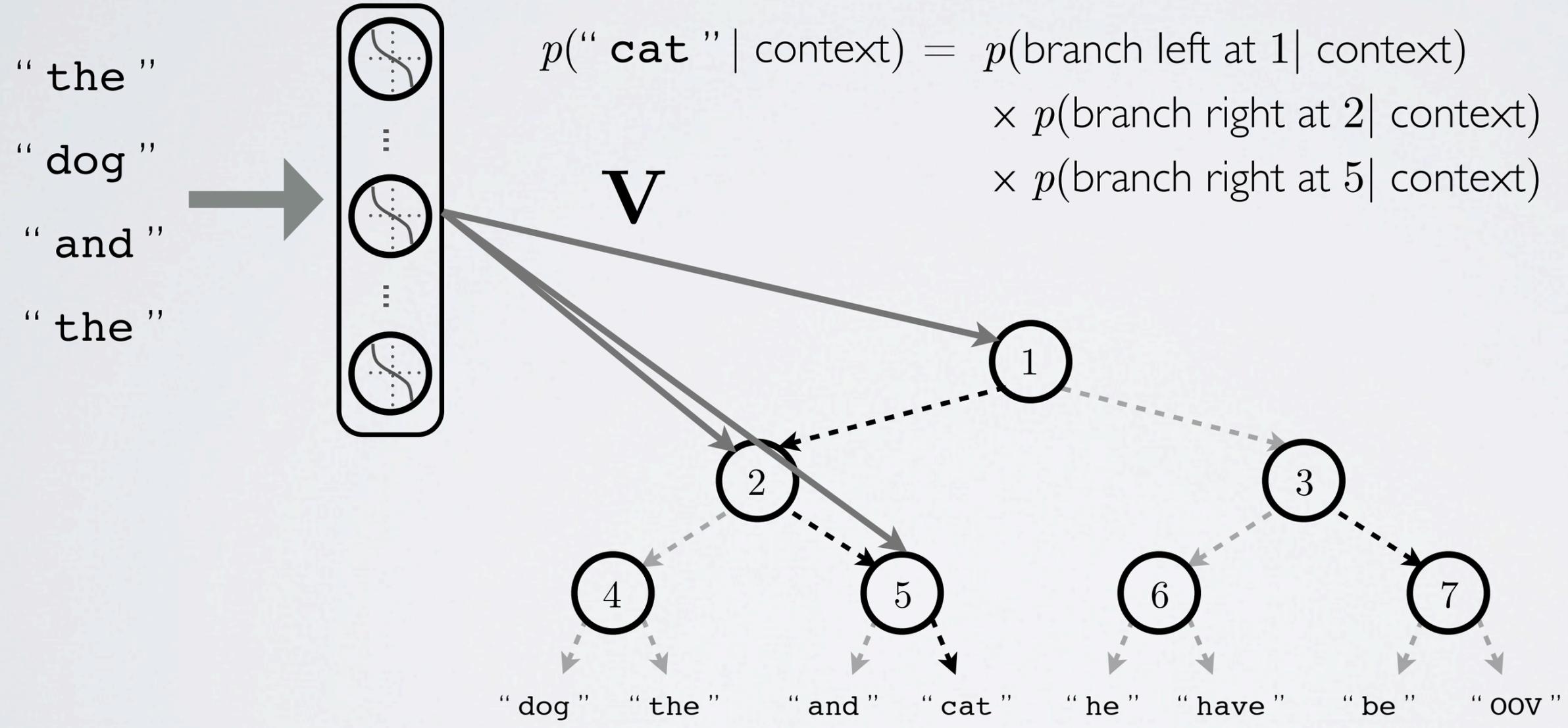
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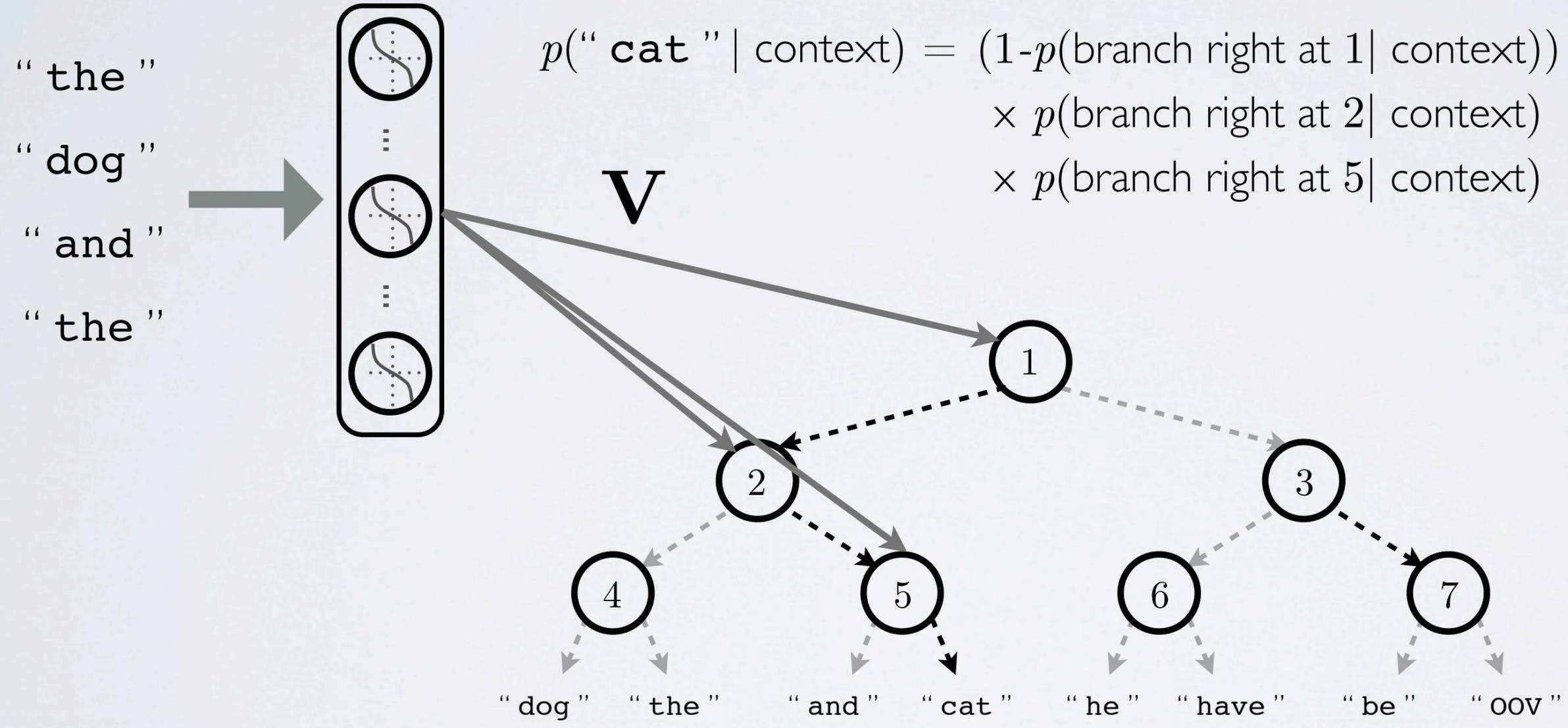
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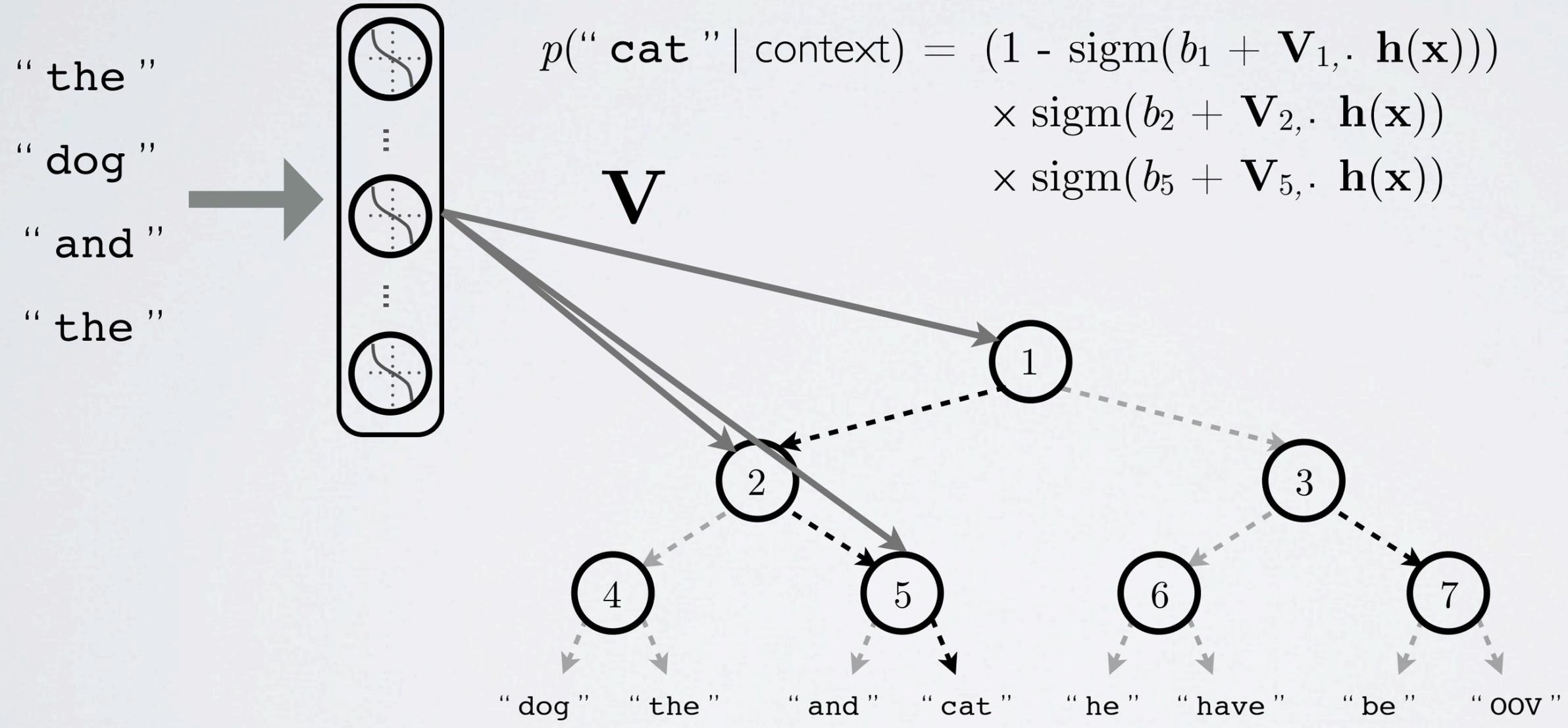
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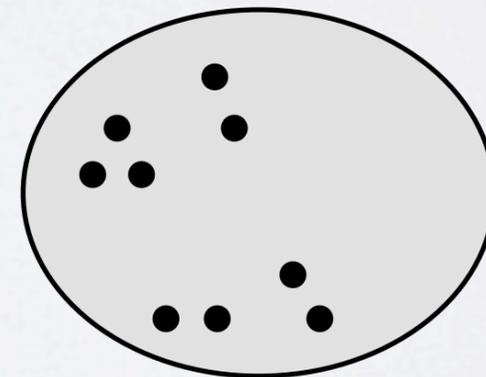
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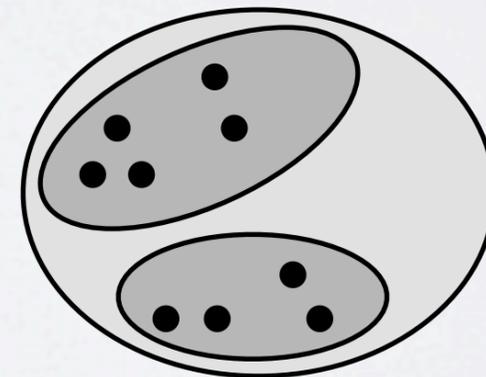
- How to define the word hierarchy
 - ▶ can use a randomly generated tree
 - this is likely to be suboptimal
 - ▶ can use existing linguistic resources, such as WordNet
 - Hierarchical Probabilistic Neural Network Language Model
Morin and Bengio, 2005
 - they report a speedup of 258x, with a slight decrease in performance
 - ▶ can learn the hierarchy using a recursive partitioning strategy
 - A Scalable Hierarchical Distributed Language Model
Mnih and Hinton, 2008
 - similar speedup factors are reported, without a performance decrease



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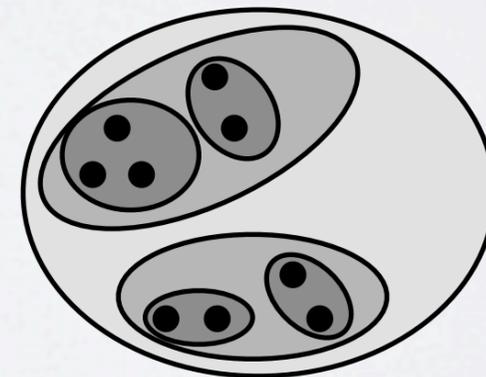
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