## Prefix Probabilities for Linear Indexed Grammars

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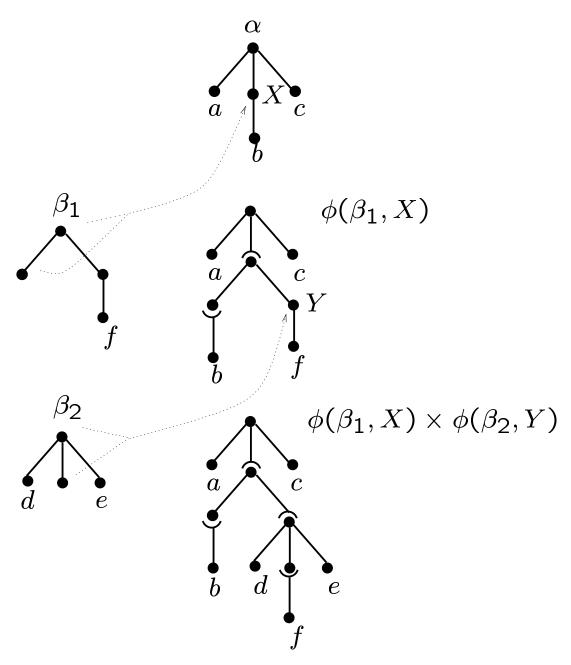
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#### **Prefix Probabilities**

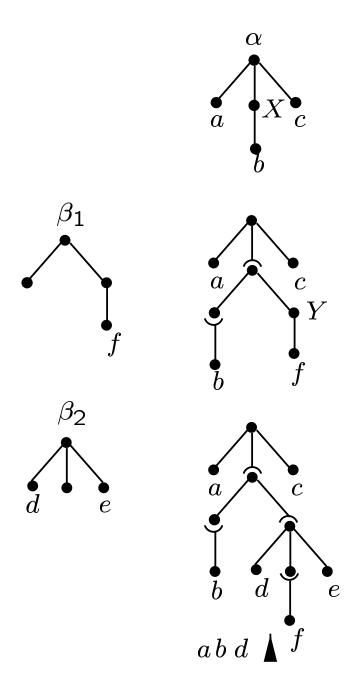
- Language model: given a string  $a_1, \ldots, a_{i-1}$ ,  $a_i$  can be any word in the vocabulary  $\Sigma$ , what is  $P(a_i \mid a_1, \ldots, a_{i-1})$ ?
- Standard techniques use trigram models:  $P(a_i \mid a_{i-2}, a_{i-1})$
- A stochastic grammar can be used by computing the prefix probability:

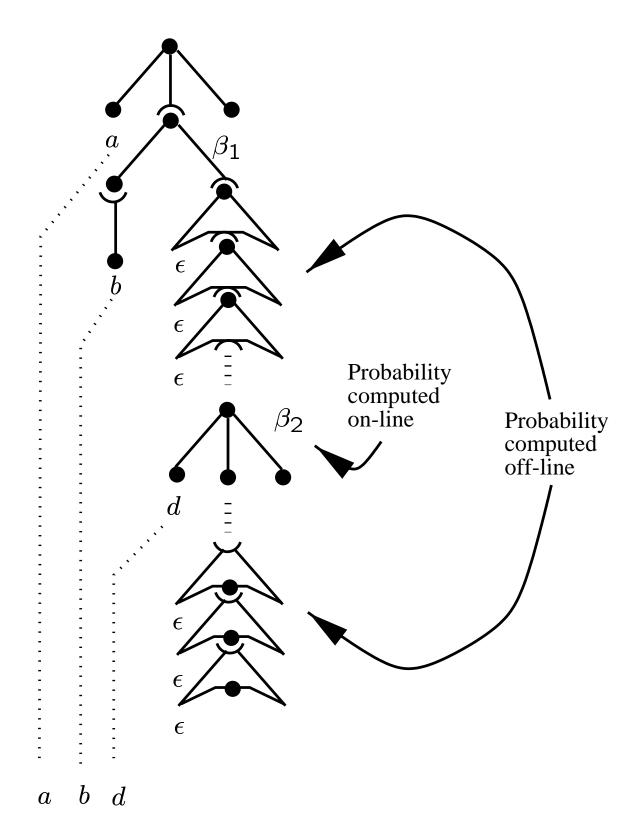
$$\sum_{w \in \Sigma^*} P(a_1, \dots, a_n w)$$

## Stochastic Tree Adjoining Grammars



### Let prefix = abd





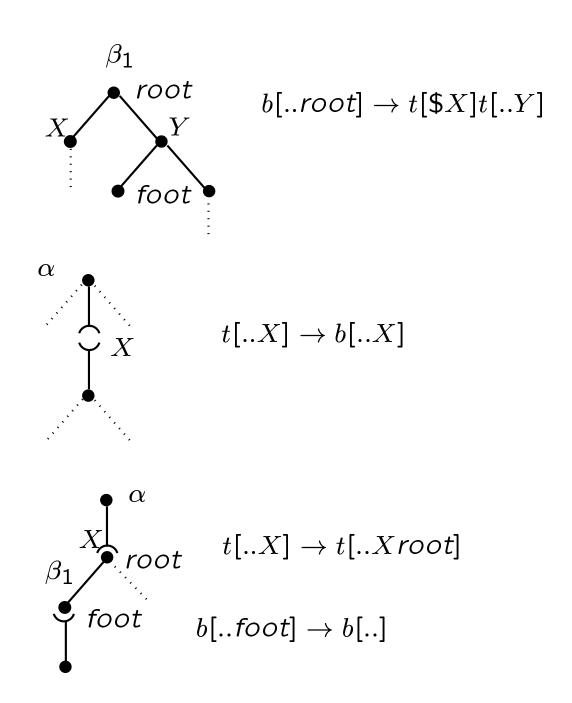
#### **Problem**

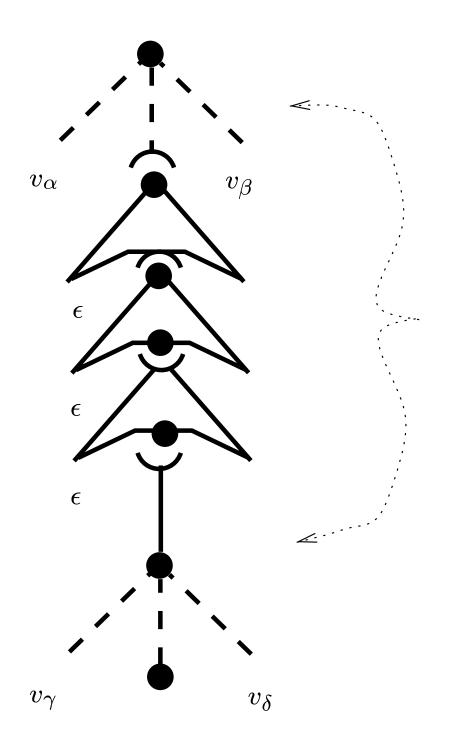
- Derivations are a combination of two kinds of subderivations:
  - 1. potentially unbounded subderivations, independent of input
  - 2. bounded subderivations, depend on input symbols
- Problem: how to partition derivations uniquely into subderivations.
- Without unique partitions, algorithm will return incorrect probabilities.

#### TAG Derivations and LIGs

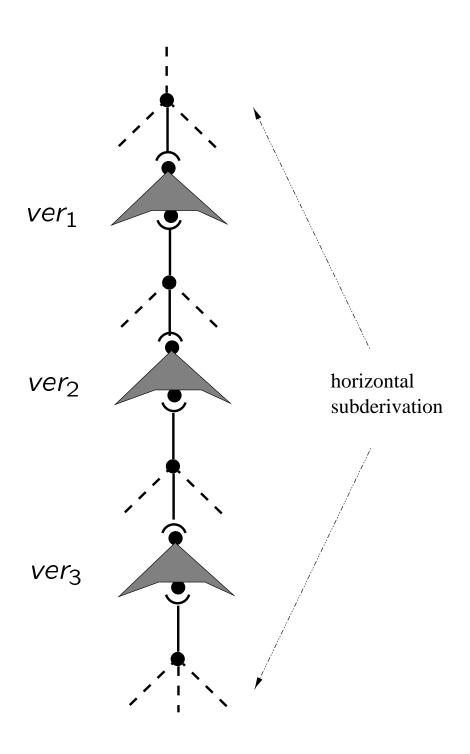
- CFGs with stack symbols, Indexed Grammars:  $A[..\eta_r] \rightarrow B[..\eta_1]C[..\eta_2]$
- Unique rhs gets stack = Linear Indexed Grammars:  $A[..\eta_r] \rightarrow B[..\eta_1]C[\$\eta_2]$
- LIGs and TAGs are weakly equivalent.
- Given a TAG, a strongly equivalent LIG can be built.
- LIGs offer a convenient way to denote TAG derivations.

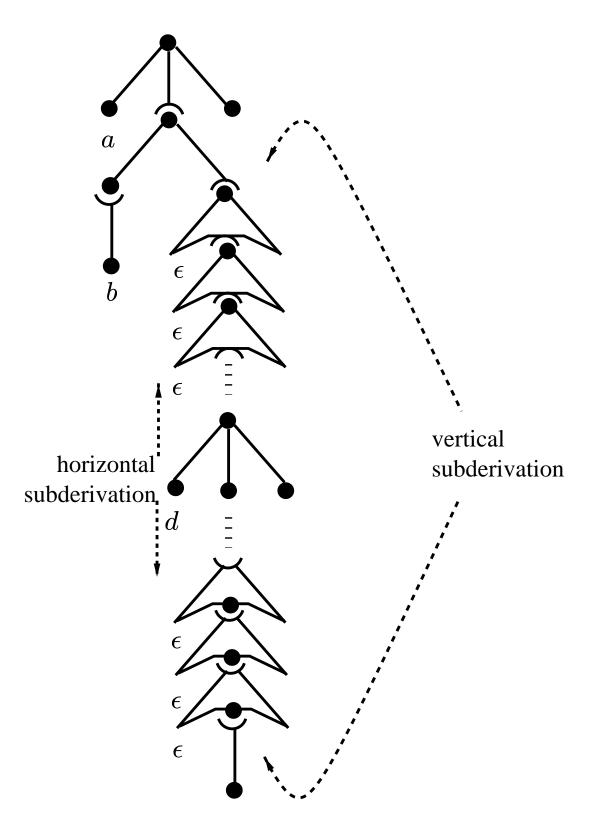
#### TAG Derivations and LIGs





# vertical subderivation





#### **Summary**

- LIGs offer a convenient way to denote TAG derivations.
- Potentially unbounded subderivations, independent of input. Probability is computed off-line.
- Bounded subderivations, depends on input symbols. Inside probability computed as prefix is recognized.
- Algorithm partitions derivations uniquely into subderivations.