

CMSC 132: OBJECT-ORIENTED PROGRAMMING II



Markov Chains, and Random Text Generation

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Random Text Generation Project

- Goal
 - Read in text
 - Generate similar semi-random text
- Approach
 1. Build DenseBag to store word frequencies
 2. Use DenseBag to build Markov chain
 3. Use Markov chain to generate semi-random text

DenseBag

- Properties
 - Like a Set
 - But can contain duplicates
- Examples
 - $\{ 1, 3, 1, 1, 3, 5 \}$
 - $\{ 1, 1, 1, 3, 3, 5 \}$
 - $\{ \text{three } 1\text{'s, two } 3\text{'s, one } 5 \}$
 - All represent same DenseBag

DenseBag<E> Operations

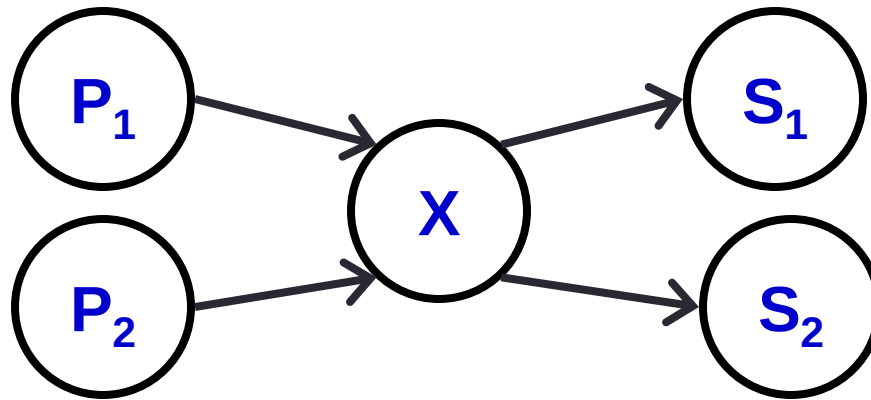
- Operations supported
 - `Set<E> getUniqueElements( )`
 - `int getCount(E e)`
 - `E choose(Random r)`
- Examples
 - Given `DenseBag<Integer> x = { 1, 1, 1, 3, 3, 5 };`
 - `x.getUniqueElements( ) → { 1, 3, 5 }`
 - `x.getCount( 1 ) → 3`
 - `x.choose(r) → 1 (50%), 3 (33%) or 5 (17%)`

DenseBag<E> Operations

- Efficiency
 - Most operations should take $O(1)$
 - If using hashing
 - `choose(Random r)` may take $O(|\text{unique items}|)$
- Iterator
 - Iterates over all elements
 - Order is undefined

Markov Chain

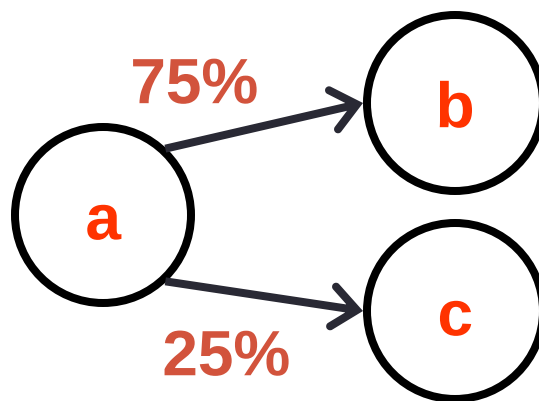
- Definition
 - A series of states with the **Markov property**
 - Where probability of future states depends only upon the present state and not on any past states
 - Example: Probability of X going to S_1 or S_2 is independent of whether P_1 or P_2 originally moved to X



- Used in
 - Statistical machine learning (artificial intelligence)

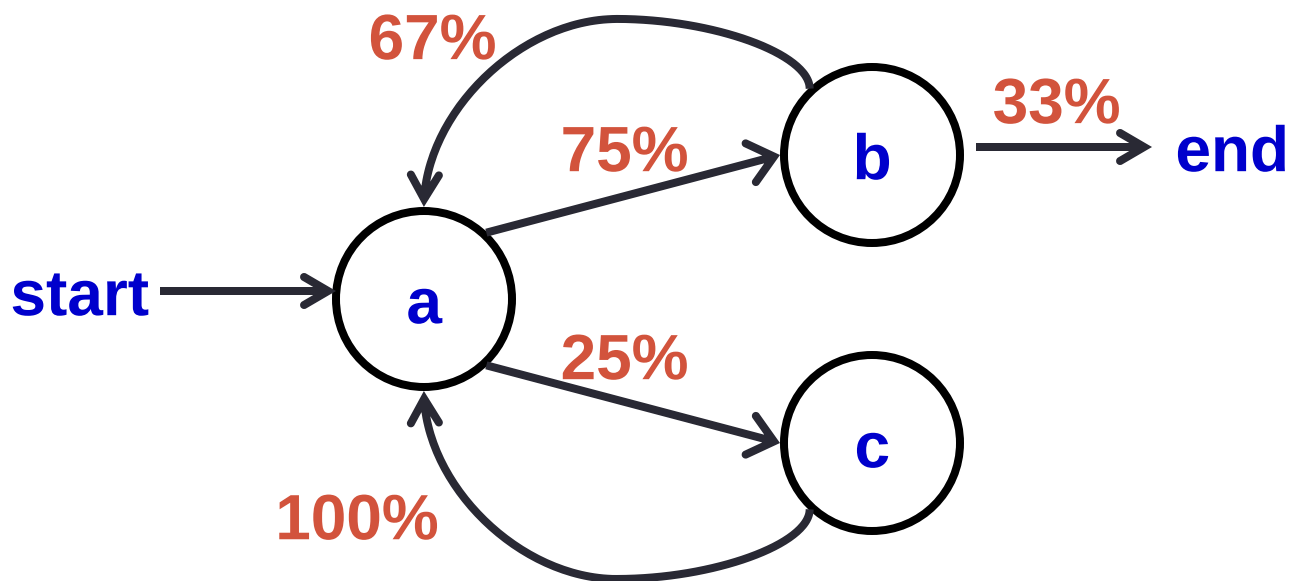
Markov Chain For Text

- Application of Markov chain
 - Represent probability of word following each word
 - Based on actual frequencies found in text
- Example
 - In the text “a b a c a b a b”
 - Word **a** is followed by **b** (75%) or **c** (25%)
 - Markov chain for words following **a**



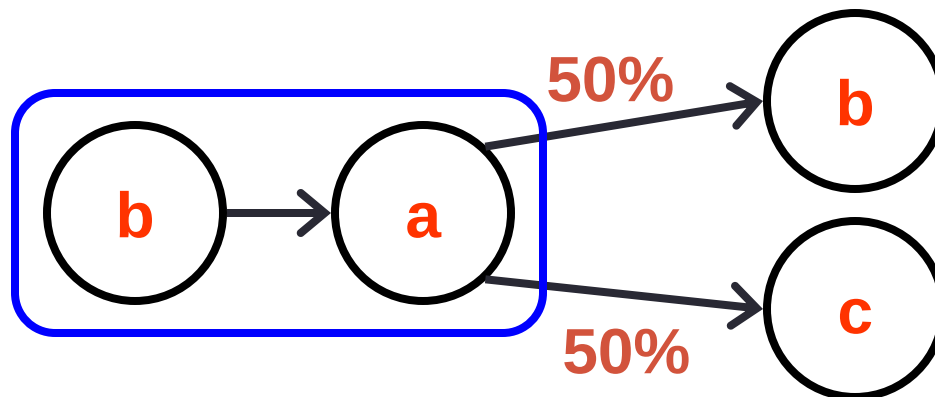
Markov Chain For Text

- Example
 - For the text “a b a c a b a b”
 - Markov chain for entire text



Higher-Order Markov Chain

- Application
 - Can represent probability of word following each **group** of words (order-**k** for **k** consecutive words)
- Example
 - In the text “**a** **b** **a** **c** **a** **b** **a** **b**”
 - Words **b** **a** are followed by **b** (50%) or **c** (50%)
 - Represent with following Markov chain

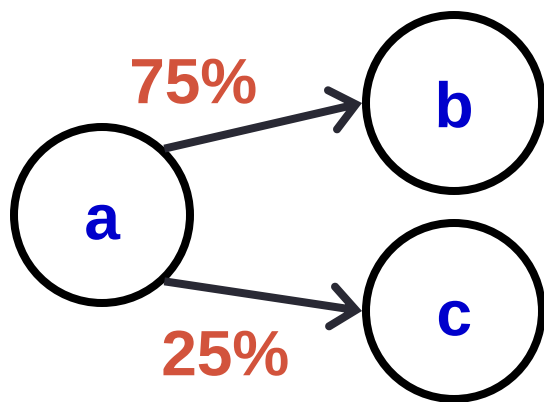


DenseBag → Markov Chain

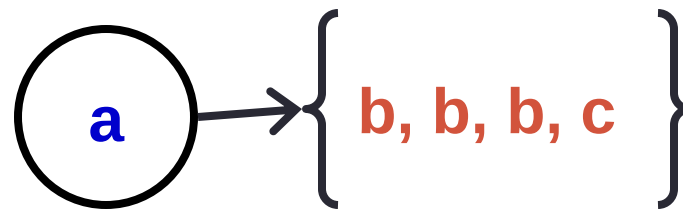
- DenseBag can represent state in Markov chain
 - Contains output in proportion to probability

- Example

- Markov state transitions



DenseBag



Markov Text Generation

- Approach (for order- n Markov text)
 1. Generate higher-order Markov chains
 - Analyze “training” text(s)
 2. Represent Markov chains as DenseBags
 3. Connect DenseBags
 - To build probabilistic transition table
 4. Use transition table to generate text

Handling Start & End of Text

1. Use empty string(s)
 - Start text generation with “”
 - End text if “” generated
 - “” $\rightarrow$ “a”
 - “”, “” $\rightarrow$ “”, “a”
 - “a” $\rightarrow$ “”
2. Augment input with <Start> & <End> markers
 - “a b a c” $\rightarrow$ “<Start> a b a c <End>”
 - Start text generation with <Start>
 - End text if <End> generated