

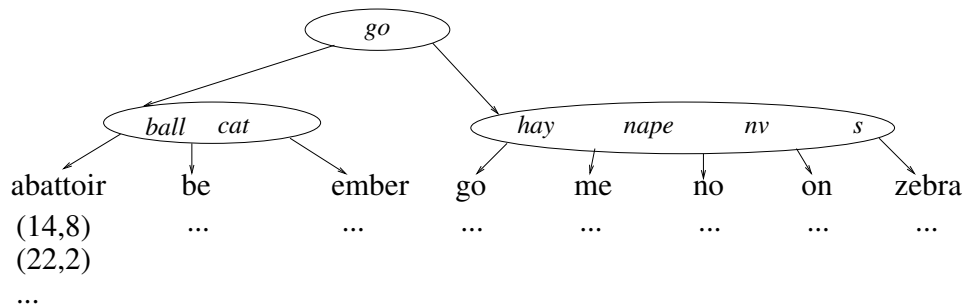
Topics: more on B-trees; retrieval based on the vector-space model, allowing the incorporation of more information regarding term frequencies.

Follow-ups to last time: If you query Google for “how tall is mount everest”, you receive a specialized response:

Mount Everest — Elevation: 8,850 M (29,035 feet) Ranked 1st
 According to http://copernicus.subdomain.de/Mount_Everest - More sources >>

But what if you query Google for “how tall is greek peak”?¹ Nothing doing.

I. Example B-tree Remember that the index is stored (in sorted order, left to right) in the leaves.



This tree is of order $t = 2$ because the root contains between 1 and $2 \cdot 2$ keys inclusive, and the other internal nodes contain between 2 and $2 \cdot 2$ (inclusive) keys. See previous lecture aid for definitions.

II. B-tree search (See that you understand from the definitions given on lecture aid from last time why this procedure works.) Suppose we want to know which documents contain the term w . Begin at the left-hand side of the root. Move rightwards among the (sorted) keys until you either *first* hit a key $k \neq w$ such that k alphabetically follows w or you hit the right end of the node; then, follow the child that you are “at”. Repeat until you hit a leaf.

III. Example data Let the vocabulary W be w_1 : cat; w_2 : dog; w_3 : news.

A three-document corpus: d : “cat news” d' : “news cat news cat news” d'' : “cat dog news dog news”

Query q : “cat dog”

IV. Normalized term-frequency vectors Let the terms be w_1, w_2, \dots, w_m . Define the *term-document frequency* $\text{freq}(w_i \in d)$ as the number of times w_i occurs in d . We then set the document vector \vec{d} for document d as follows:

$$\vec{d} = \left(\frac{\text{freq}(w_1 \in d)}{N(d)}, \frac{\text{freq}(w_2 \in d)}{N(d)}, \dots, \frac{\text{freq}(w_m \in d)}{N(d)} \right)$$

where $N(d) = \sqrt{\sum_{i=1}^m \text{freq}(w_i \in d)^2}$ is the *length-normalization factor*.

¹2100 feet.