

Lecture 20

#### Great Theoretical Ideas In Computer Science

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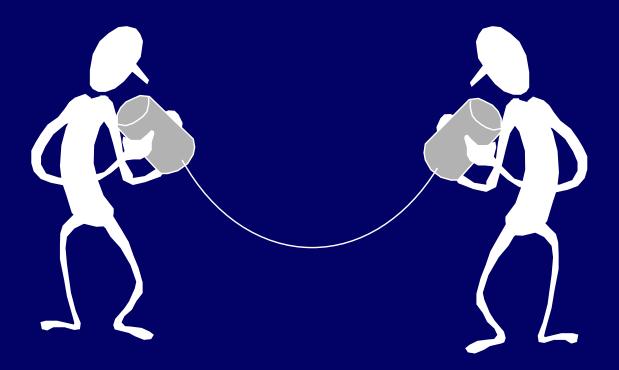
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CS 15-251

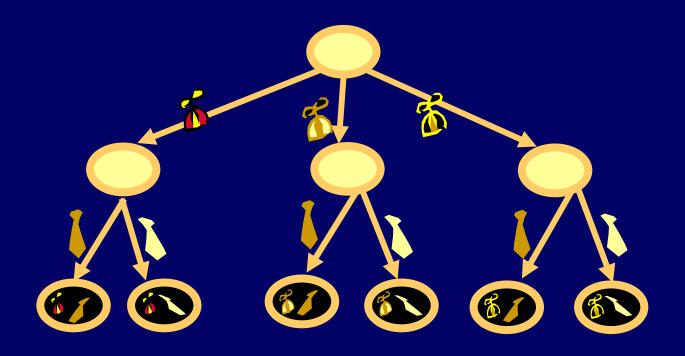
Spring 2004

Carnegie Mellon University

## Decision Trees and Information: A Question of Bits

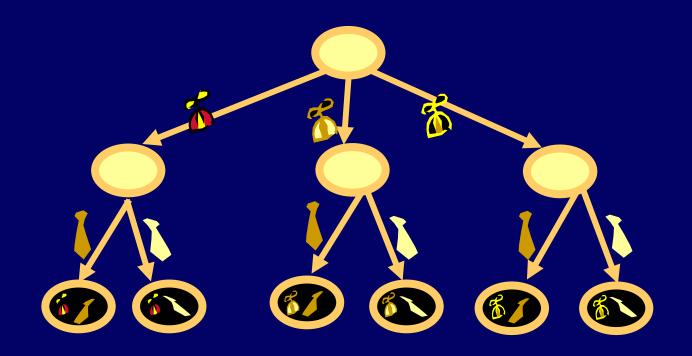


### Choice Tree



A choice tree is a rooted, directed tree with an object called a "choice" associated with each edge and a label on each leaf.

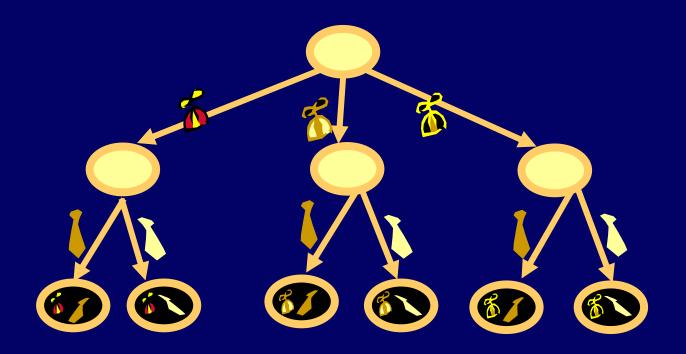
## Choice Tree Representation of S



We satisfy these two conditions:

- · Each leaf label is in S
- Each element from S on exactly one leaf.

## Question Tree Representation of S

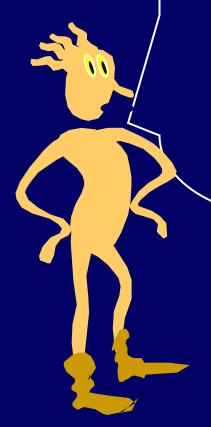


I am thinking of an outfit.

Ask me questions until you know which one.

What color is the beanie? What color is the tie?

When a question tree has at most 2 choices at each node, we will call it a decision tree, or a decision strategy.



Note: Nodes with one choices represent stupid questions, but we allow stupid questions.

## 20 Questions

S = set of all English nouns

#### Game:

I am thinking of an element of S. You may ask up to 20 YES/NO questions.

What is a question strategy for this game?

## 20 Questions

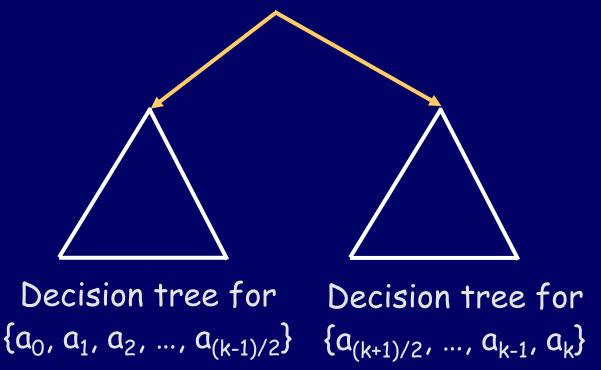
Suppose 
$$S = \{a_0, a_1, a_2, ..., a_k\}$$

Binary search on S.

First question will be: "Is the word in  $\{a_0, a_1, a_2, ..., a_{(k-1)/2}\}$ ?"

## 20 Questions Decision Tree Representation

A decision tree with depth at most 20, which has the elements of S on the leaves.



## Decision Tree Representation

#### Theorem:

## Another way to look at it

If you are thinking of the noun a<sub>m</sub> in S We are asking about each bit of index m Is the leftmost bit of m 0? Is the next bit of m 0?

•••

Theorem: The binary-search decision-tree for 
$$S = \{ a_0, a_1, a_2, ..., a_k \}$$
 has depth  $|k| = \lfloor \log k \rfloor + 1$ 

### A lower bound

Theorem: No decision tree for S can have depth d  $< \lfloor \log k \rfloor + 1$ .

#### Proof:

Any depth d tree can have at most  $2^d$  leaves. But  $d < \lfloor \log k \rfloor + 1 \Rightarrow$  number of leaves  $2^d < (k+1)$  Hence some element of S is not a leaf.

## Tight bounds!

The optimal-depth decision tree for any set S with (k+1) elements has depth

$$\lfloor \log k \rfloor + 1 = |k|$$

## Prefix-free Set

Let T be a subset of  $\{0,1\}^*$ .

#### Definition:

T is prefix-free if for any distinct  $x,y \in T$ , if |x| < |y|, then x is not a prefix of y

#### Example:

{000, 001, 1, 01} is prefix-free {0, 01, 10, 11, 101} is not.

## Prefix-free Code for S

Let S be any set.

Definition: A prefix-free code for S is a prefix-free set T and a 1-1 "encoding" function f: S -> T.

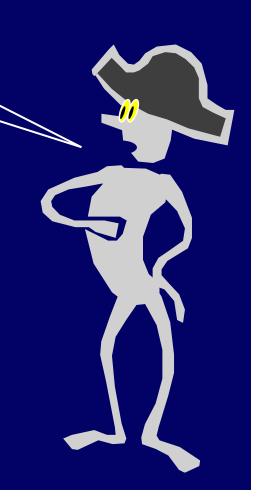
The inverse function f<sup>-1</sup> is called the "decoding function".

```
Example: S = {apple, orange, mango}.

T = {0, 110, 1111}.

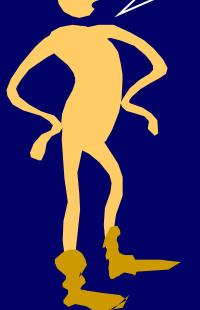
f(apple) = 0, f(orange) = 1111, f(mango) = 110.
```

## What is so cool about prefix-free codes?



# Sending sequences of elements of S over a communications channel





Let T be prefix-free and f be an encoding function. Wish to send  $\langle x_1, x_2, x_3, ... \rangle$ 

Sender: sends  $f(x_1) f(x_2) f(x_3)...$ 

Receiver: breaks bit stream into elements of T and decodes using f<sup>-1</sup>

## Sending info on a channel

```
Example: S = {apple, orange, mango}.
 T = \{0, 110, 1111\}.
 f(apple) = 0, f(orange) = 1111, f(mango) = 110.
If we see
      00011011111100...
we know it must be
      0 0 0 110 1111 110 0 ...
and hence
      apple apple mango orange mango apple ...
```

## Morse Code is not Prefix-free!

SOS encodes as ...---...

A	F	K	P	U	Z
			Q		
C	Н	M	R	W	
D	I	N	5	X	
Ε.	J	0	T <b>-</b>	У	

## Morse Code is not Prefix-free!

SOS encodes as ...---...

Could decode as: IAMIE

## Unless you use pauses

SOS encodes as ... --- ...

```
A.- F..-. K-.- P.--. U..- Z--..
B-... G--. L.-.. Q--.- V...-
C-.-. H.... M-- R.-. W.--
D-.. I.. N-. S... X-..-
E. J.--- O--- T- Y-.--
```

Prefix-free codes are also called "self-delimiting" codes.



## Representing prefix-free codes

A = 100

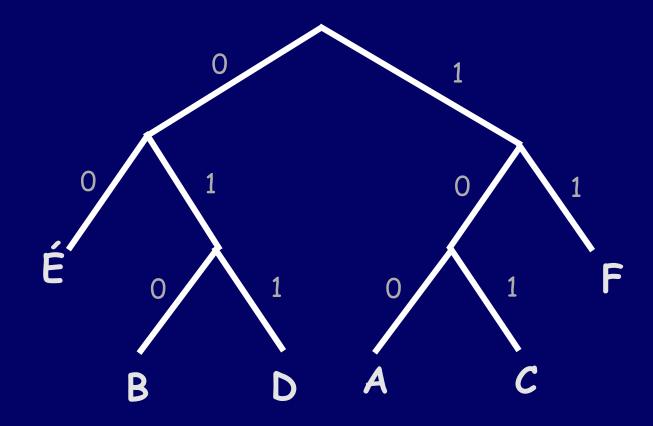
B = 010

C = 101

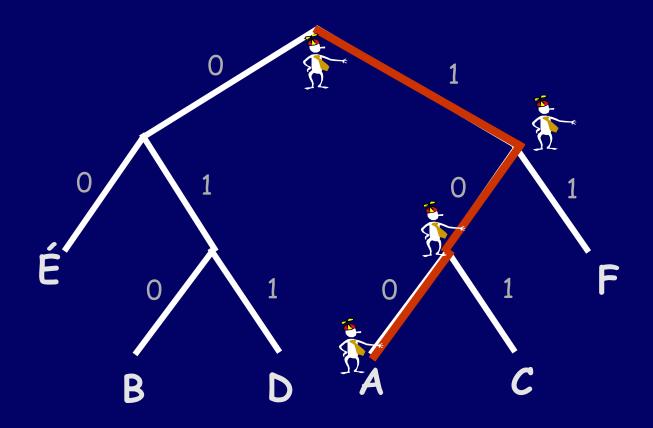
D = 011

É = 00

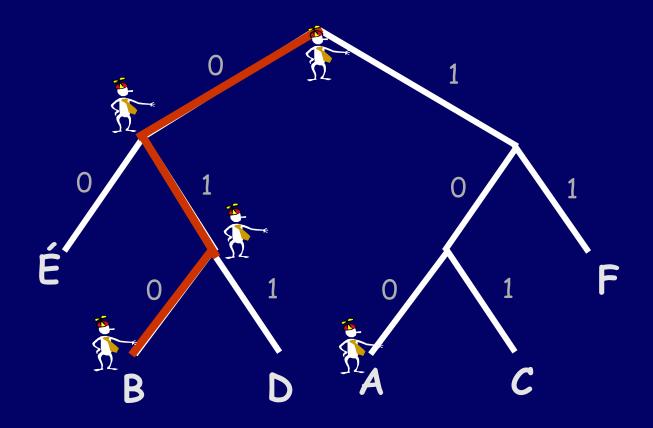
F = 11



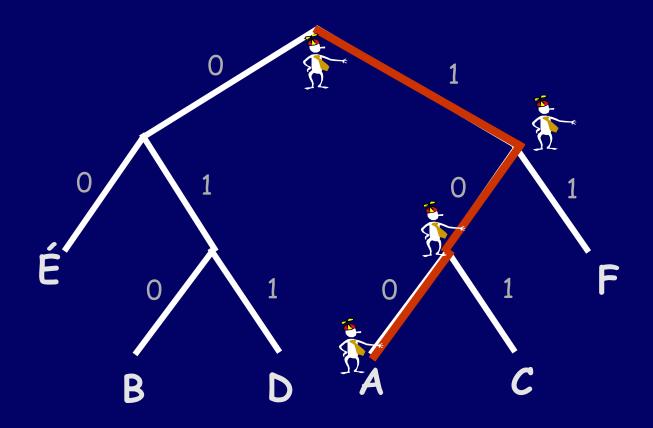
"CAFÉ" would encode as 1011001100 How do we decode 1011001100 (fast)?



can decode as:

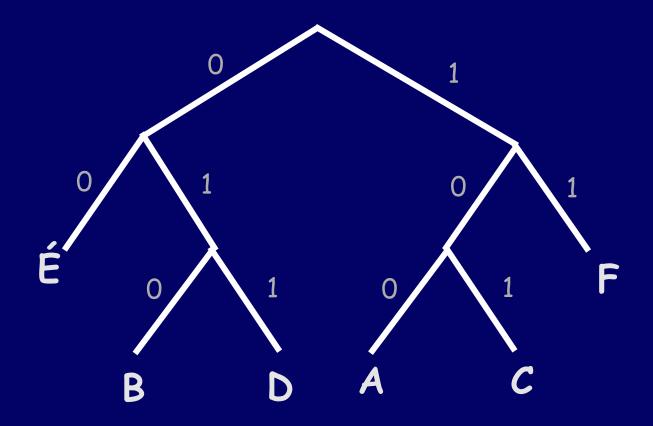


can decode as: A



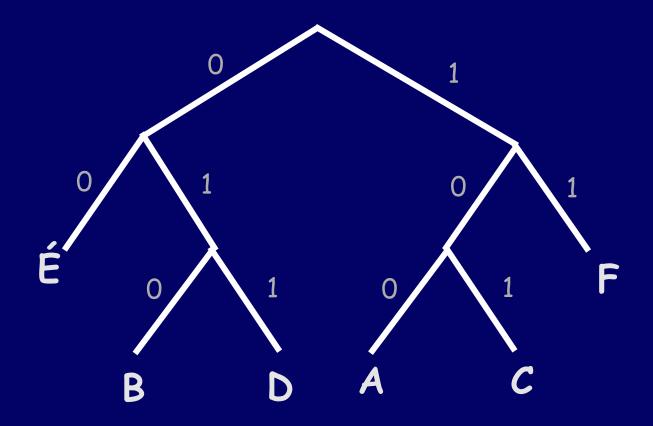
If you see: 100<mark>010</mark>1000111011001100

can decode as: AB

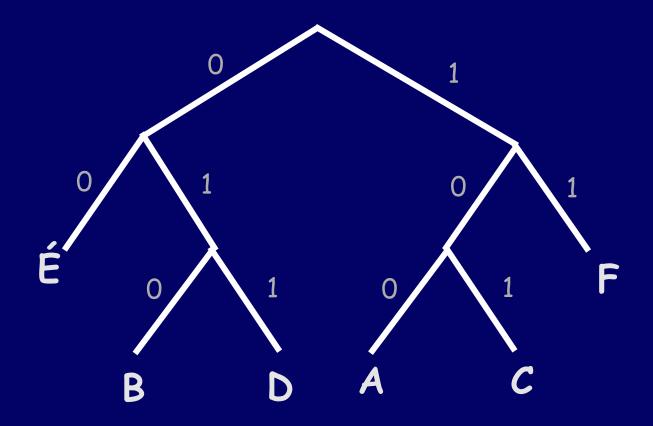


If you see: 100010<mark>100</mark>0111011001100

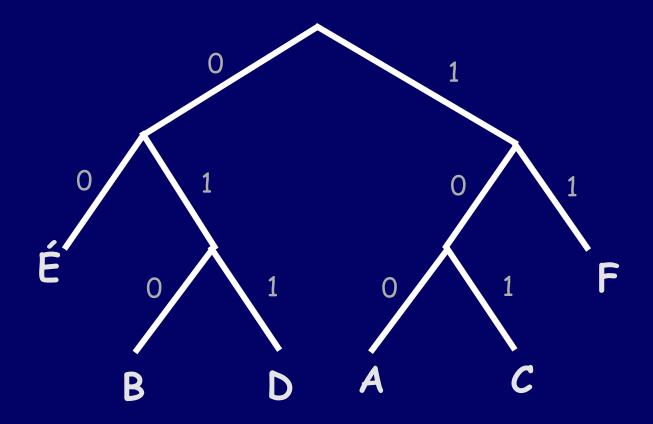
can decode as: ABA



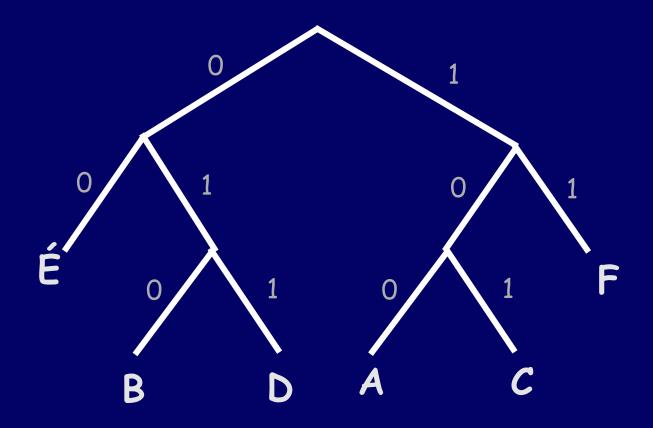
can decode as: ABAD



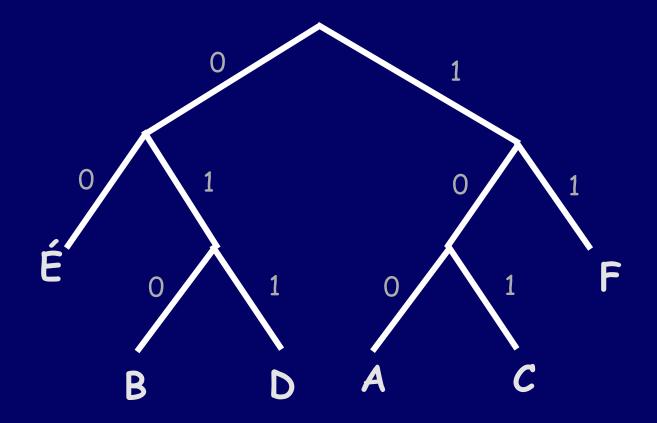
can decode as: ABADC



can decode as: ABADCA



can decode as: ABADCAF



can decode as: ABADCAFÉ

Prefix-free codes are yet another representation of a decision tree.

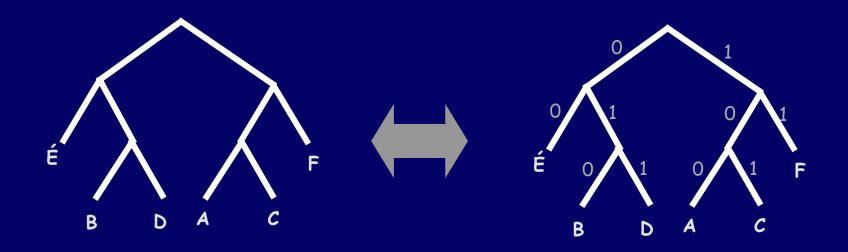


#### Theorem:

S has a decision tree of depth d

if and only if

S has a prefix-free code with all codewords bounded by length d



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S has a decision tree of depth d if and only if

S has a prefix-free code with all codewords bounded by length d

## Extends to infinite sets

Let S is a subset of  $\Sigma^*$ 

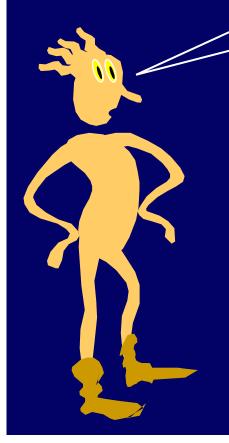
#### Theorem:

S has a decision tree where all length n elements of S have depth  $\leq D(n)$ 

if and only if

S has a prefix-free code where all length n strings in S have encodings of length  $\leq D(n)$ 

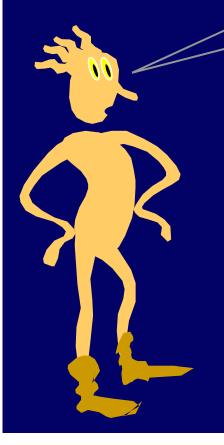
I am thinking of some natural number k - ask me YES/NO questions in order to determine k.



Let d(k) be the number of questions that you ask when I am thinking of k.

Let  $D(n) = max \{ d(j) \text{ over } n\text{-bit numbers } j \}$ .

I am thinking of some natural number k ask me YES/NO questions in order to determine k.



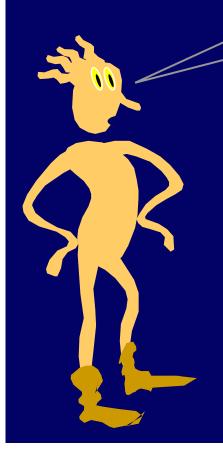
Lousy strategy: Is it 0? 1? 2? 3? ...

d(k) = k+1

 $D(n) = 2^{n+1}$  since  $2^{n+1} - 1$  uses only n bits.

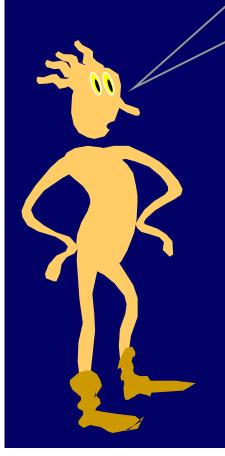
Effort is exponential in length of k !!!

I am thinking of some natural number k - ask me YES/NO questions in order to determine k.



What is an efficient question strategy?

## I am thinking of some natural number k...



Does k have length 1? NO Does k have length 2? NO Does k have length 3? NO

•••

Does k have length n? YES

Do binary search on strings of length n.

$$d(k) = |k| + |k|$$
  
= 2 ( | log k | + 1 )

D(n) = 2n

#### Size First/Binary Search

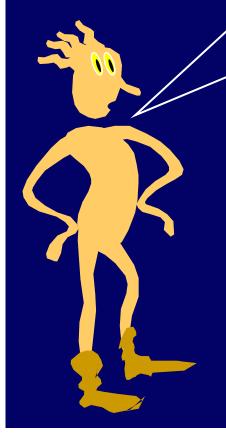
Does k have length 1? NO Does k have length 2? NO Does k have length 3? NO

•••

Does k have length n? YES

Do binary search on strings of length n.

# What prefix-free code corresponds to the Size First / Binary Search decision strategy?



f(k) = (|k| - 1) zeros, followed by 1, and then by the binary representation of k

|f(k)| = 2 |k|

What prefix-free code corresponds to the Size First / Binary Search decision strategy?



Or,

length of k in unary  $\Rightarrow$  |k| bits k in binary  $\Rightarrow$  |k| bits

## Another way to look at f

k = 27 = 11011, and hence |k| = 5

f(k) = 00001 11011

## Another way to look at f

$$k = 27 = 11011$$
, and hence  $|k| = 5$ 

q(k) = 0101000111

11011

0101000111

Another way to look at the function g:

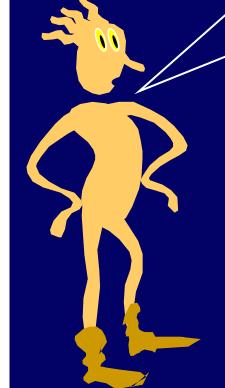
$$g(final 0) \rightarrow 10$$
  $g(all other 0's) \rightarrow 00$   
 $g(final 1) \rightarrow 11$   $g(all other 1's) \rightarrow 01$ 

"Fat Binary"  $\Leftrightarrow$  Size First/Binary Search strategy

Is it possible to beat 2n questions to find a number of length n?

## Look at the prefix-free code...

Any obvious improvement suggest itself here?



the fat-binary map f concatenates

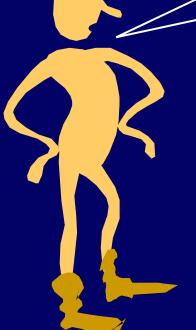
length of k incurry  $\Rightarrow |k|$  bits k in binary  $\Rightarrow |k|$  bits

fat binary!

In fat-binary,  $D(n) \le 2n$ Now  $D(n) \le n + 2 (\lfloor \log n \rfloor + 1)$ 

Can you do better?





better-than-Fat-Binary-code(k) concatenates

length of k in fat binary  $\Rightarrow 2||k||$  bits k in binary  $\Rightarrow |k|$  bits

Hey, wait!

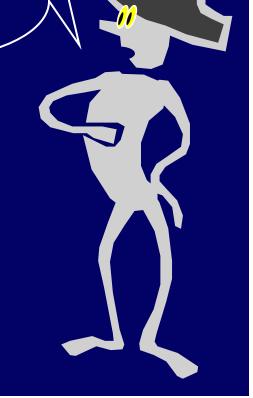
In a better prefix-free code

RecursiveCode(k) concatenates
RecursiveCode(|k|) & k in binary

better-t-better-thanFB

better than Fat Dinary code

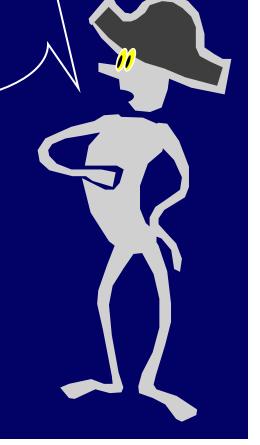
better-t-FB ||k|| + 2|||k||| |k| in fat binary  $\Rightarrow 2||k||$  bits |k| in binary  $\Rightarrow |k|$  bits

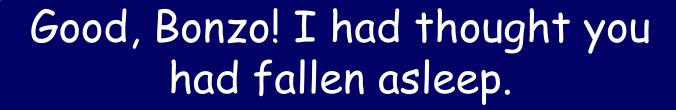


Oh, I need to remember how many levels of recursion r(k)

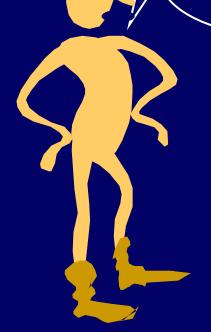
In the final code F(k) = F(r(k)). RecursiveCode(k)

$$r(k) = log* k$$









Maybe I can do better...

Can I get a prefix code for k with length ≈ log k?

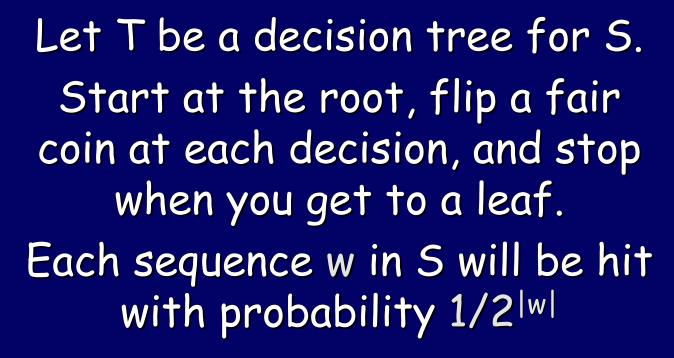


#### No!

Let me tell you why length ≈ log k is not possible

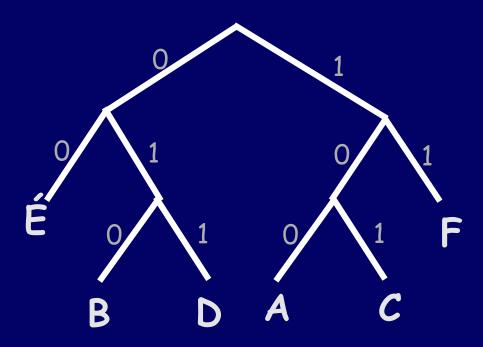


Decision trees have a natural probabilistic interpretation.





#### Random walk down the tree



Each sequence w in S will be hit with probability 1/2 |w|

Hence,  $Pr(F) = \frac{1}{4}$ , Pr(A) = 1/8, Pr(C) = 1/8, ...

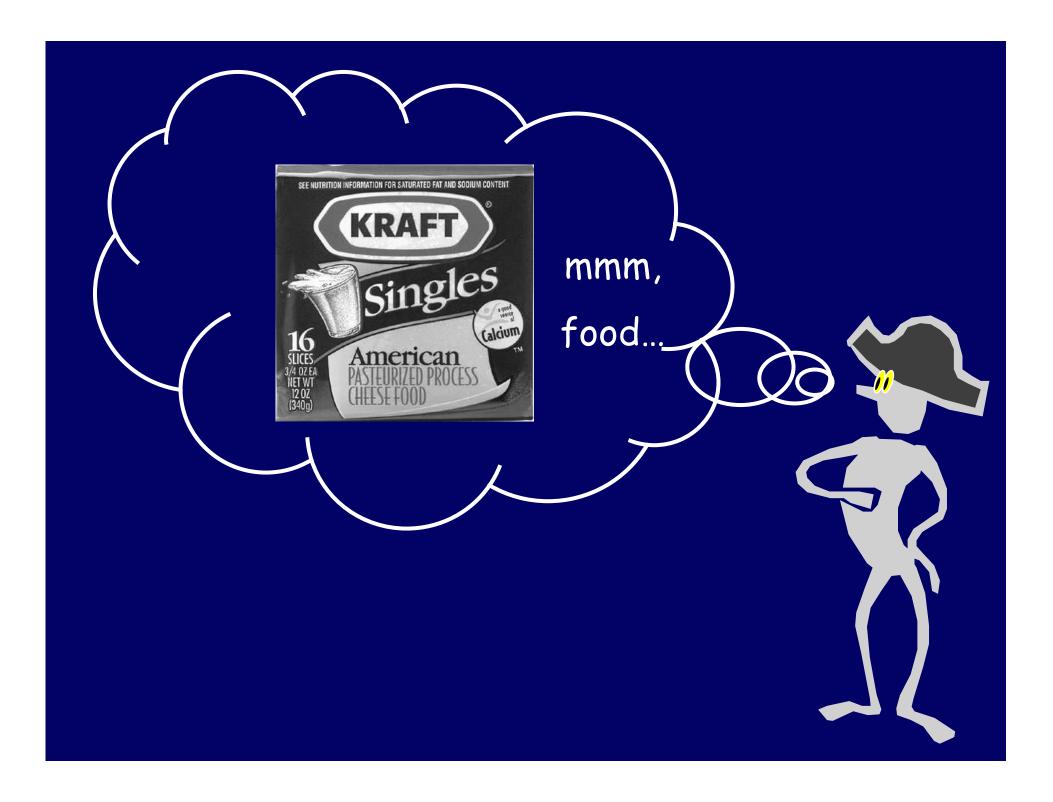


The probability that some element in S is hit by a random walk down from the root is

 $|\sum_{w \in S} 1/2^{|w|} \leq 1$ 

Kraft Inequality

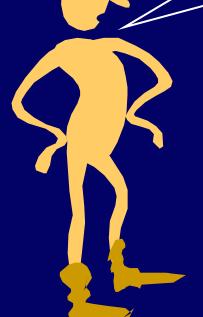




Kraft Inequality:

$$\sum_{w \in S} 1/2^{|w|} \leq 1$$





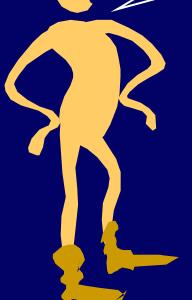
Fat Binary: f(k) has  $2|k| \approx 2 \log k$  bits

$$\sum_{\mathsf{k}\in\mathbb{N}} 1/\mathsf{k}^2 \leq \sum_{\mathsf{k}\in\mathbb{N}} \frac{1}{2} |\mathsf{f}(\mathsf{k})| \leq 1$$

Kraft Inequality:

$$\sum_{w \in S} 1/2^{|w|} \leq 1$$





Better Code: f(k) has |k| + 2||k|| bits

$$\sum_{k \in \mathbb{N}} 1/(k (\log k)^2) \leq 1$$

## Kraft Inequality:

$$\sum_{w \in S} 1/2^{|w|} \leq 1$$





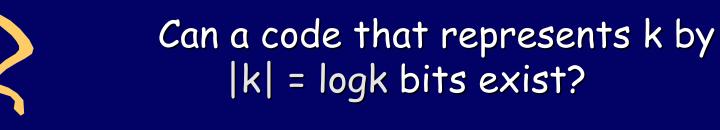
Ladder Code: k is represented by | |k| + ||k|| + |||k||| + ... bits

$$\sum_{k \in \mathbb{N}} 1/(k \log k \log \log k ...) \leq 1$$

Kraft Inequality:

$$\sum_{w \in S} 1/2^{|w|} \leq 1$$





No, since  $\sum_{k \in \mathbb{N}} 1/k$  diverges !! So you can't get log n, Bonzo...



## Back to compressing words

The optimal-depth decision tree for any set S with (k+1) elements has depth  $\lfloor \log k \rfloor + 1$ 



The optimal prefix-free code for A-Z + "space" has length  $\lfloor \log 26 \rfloor + 1 = 5$ 

## English Letter Frequencies

But in English, different letters occur with different *frequencies*.

A 8.1%	F 2.3%	K .79%	P 1.6%	U 2.8%	Z .04%
B 1.4%	G 2.1%	L 3.7%	Q .11%	V .86%	
C 2.3%	H 6.6%	M 2.6%	R 6.2%	W 2.4%	
D 4.7%	I 6.8%	N 7.1%	5 6.3%	X .11%	
E 12%	J .11%	07.7%	T 9.0%	y 2.0%	

ETAONIHSRDLUMWCFGYPBVKQXJZ

## short encodings!

Why should we try to minimize the maximum length of a codeword?

If encoding A-Z, we will be happy if the "average codeword" is short.

#### Morse Code

```
      A.-
      F..-.
      K-.-
      P.--.
      U..-
      Z--..

      B-...
      G--.
      L.-..
      Q--.-
      V...-

      C-.-.
      H...
      M--
      R.-.
      W.--

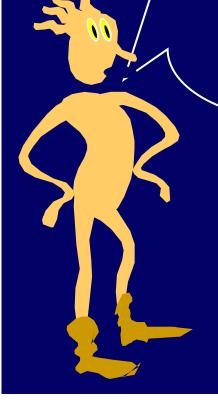
      D-..
      I..
      N-.
      S...
      X-..-

      E.
      J.---
      O---
      T-
      Y-.--
```

#### ETAONIHSRDLUMWCFGYPBVKQXJZ

Given frequencies for A-Z, what is the optimal prefix-free encoding of the alphabet?

I.e., one that minimizes the average code length



## Huffman Codes: Optimal Prefix-free Codes Relative to a Given Distribution

Here is a Huffman code based on the English letter frequencies given earlier:

A 1011	F 101001	K 10101000	P 111000	U 00100
B 111001	G 101000	L 11101	Q 1010100100	V 1010101
C 01010	H 1100	M 00101	R 0011	W 01011
D 0100	I 1111	N 1000	S 1101	X 1010100101
E 000	J 1010100110	O 1001	T 011	У 101011
				Z 1010100111

## But we are talking English!

Huffman coding uses only letter frequencies.

But we can use correlations! E.g., Q is almost always followed by U... Can we use this?

Yes...
Let us see how these correlations help

Randomly generated letters from A-Z, space not using the frequencies at all:

XFOML RXKHRJFFJUJ ALPWXFWJXYJ
FFJEYVJCQSGHYD QPAAMKBZAACIBZLKJQD

Using only single character frequencies:

OCRO HLO RGWR NMIELWIS EU LL NBNESEBYA TH EEI ALHENHTTPA OOBTTVA NAH BRL

Each letter depends on the previous letter:

ON IE ANTSOUTINYS ARE T INCTORE ST BE S
DEAMY ACHIN D ILONASIVE TUCOOWE AT
TEASONARE FUSO TIZIN ANDY TOBE SEACE CTISBE

Each letter depends on 2 previous letters:

IN NO IST LAT WHEY CRATICT FROURE BIRS
GROCID PONDENOME OF DEMONSTURES OF THE
REPTAGIN IS REGOACTIONA OF CRE

Each letter depends on 3 previous letters:

THE GENERATED JOB PROVIDUAL BETTER TRAND
THE DISPLAYED CODE, ABOVERY UPONDULTS WELL
THE CODERST IN THESTICAL IT DO HOCK
BOTHEMERG.

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#### References

The Mathematical Theory of Communication, by C. Shannon and W. Weaver

Elements of Information Theory, by T. Cover and J. Thomas