

# Information Theory and Networks

## Lecture 10: Sampling with Fair Coins

Matthew Roughan

`<matthew.roughan@adelaide.edu.au>`

[http://www.maths.adelaide.edu.au/matthew.roughan/  
Lecture\\_notes/InformationTheory/](http://www.maths.adelaide.edu.au/matthew.roughan/Lecture_notes/InformationTheory/)

School of Mathematical Sciences,  
University of Adelaide

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# Part I

## Sampling with Fair Coins

USA Today has come out with a new survey: Apparently three out of four people make up 75 percent of the population.

*David Letterman*

## Problem

*Imagine you have a fair coin, but you want to sample from an arbitrary distribution, how would you do it?*

# Example 1

Example: use a sequence of fair coin tosses to generate a random variable  $X$  with PMF

$$X = \begin{cases} a, & \text{with probability } 1/2, \\ b, & \text{with probability } 1/4, \\ c, & \text{with probability } 1/4, \end{cases}$$

# Example 1

Obvious solution:

- ① Toss the coin once:
  - ① If its a H, then  $X = a$
  - ② If its a T, toss it again
    - ① If its a H, then  $X = b$
    - ② If its a T, then  $X = c$

$$X = \begin{cases} a, & \text{with probability } 1/2, \\ b, & \text{with probability } 1/4, \\ c, & \text{with probability } 1/4, \end{cases}$$

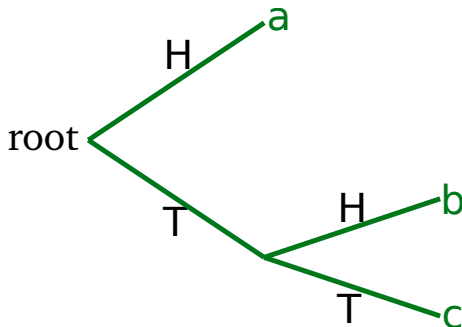
# General Problem

- We want to generate a random variable  $X \in \Omega = \{1, 2, \dots, m\}$ 
  - ▶  $X$  has PMF  $\{p_1, p_2, \dots, p_m\}$
- We have a series of (independent) fair coin tosses  $Z_1, Z_2, \dots$ 
  - ▶ let  $T$  denote the number of coin tosses (which is potentially a RV)
  - ▶ we'd like methods that minimise  $E[T]$

## Example 1

First thing to note is that the coin tosses define a binary tree:  
For example, given the solution for

$$X = \begin{cases} a, & \text{with probability } 1/2, \\ b, & \text{with probability } 1/4, \\ c, & \text{with probability } 1/4, \end{cases}$$





So consider the tree we want to generate:

- The tree should be complete: every node should be a leaf, or have two descendents.
  - ▶ i.e., at a node, we stop, or toss another coin
  - ▶ probability of a leaf at depth  $k$  is  $2^{-k}$
- The leaves correspond to outcomes for  $X$ 
  - ▶ more than one leaf could be labelled with the same outcome
  - ▶ the total probability of all leaves with the same outcome  $i$  should be  $p_i$
- The tree could be infinite
- There are multiple possible trees for some problems
  - ▶ lets aim for the most efficient
  - ▶ minimise the expected depth

## Problem

*How could we go about designing such a tree for a **dyadic** distribution (one whose probabilities are powers of two)?*

*Is it related to entropy?*

Solution: use the Huffman tree we get from treating it like a coding problem.

Why?

- Tree generates dyadic probabilities
- Probability of leaf for code  $k$  is  $2^{-\ell_k}$  where  $\ell_k$  is the length of the code (depth of the tree)
- Earlier we showed that for dyadic probabilities the optimal code lengths were

$$\ell_k = \left\lceil \log_D \left( \frac{1}{p_k} \right) \right\rceil$$

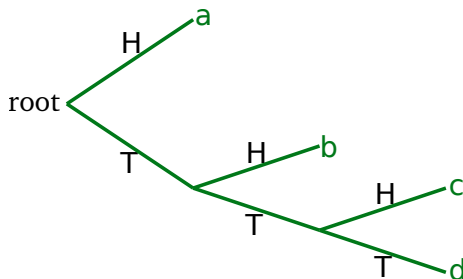
For dyadic probabilities, this gives integer lengths.

- Huffman code tree generates optimal codes
- So the probabilities the Huffman code generates are the same as the ones we need
  - ▶ and the expected number of coin tosses will be minimised
  - ▶ the expected number will be the entropy

## Example 2

$$X = \begin{cases} a, & \text{with probability } 1/2, & \text{has code } 0 \\ b, & \text{with probability } 1/4, & \text{has code } 10 \\ c, & \text{with probability } 1/8, & \text{has code } 110 \\ d, & \text{with probability } 1/8, & \text{has code } 111 \end{cases}$$

Code tree as a set of trials:  $0 = H, 1 = T$



## Problem

*What about non-dyadic probabilities?*

## Solution:

- Break probabilities into dyadic atoms
  - ▶ Write out the probability in binary (decimal) notation
  - ▶ You may need to approximate at some point
- Combine back to the original probabilities by applying the same label to the appropriate leaves.
  - ▶ sum leaves with same labels

## Example 3

$$X = \begin{cases} a, & \text{with probability } 1/3 \\ b, & \text{with probability } 2/3 \end{cases}$$

Binary expansions:

$$\begin{array}{l} \frac{2}{3} = 0.101010101\dots = 2^{-1} + 2^{-3} + \dots \\ \frac{1}{3} = 0.010101010\dots = 2^{-2} + 2^{-4} + \dots \end{array}$$

So the atoms we need are

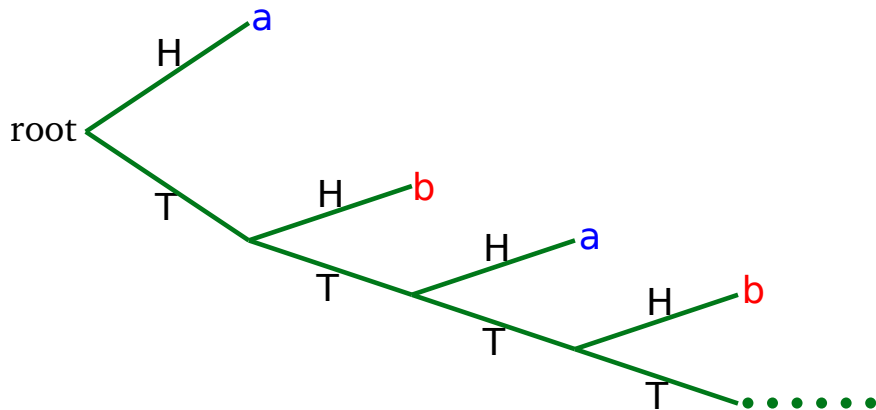
$$(p_1, p_2, \dots) = \left( \frac{1}{2}, \frac{1}{4}, \frac{1}{8}, \frac{1}{16}, \dots \right)$$

and colour indicates the label.

## Example 3

We need a Huffman code tree for the PMF

$$(p_1, p_2, \dots) = \left( \frac{1}{2}, \frac{1}{4}, \frac{1}{8}, \frac{1}{16}, \dots \right)$$





## Problem

*What if you don't even know if your coin is fair?*

Solution: You can always get a fair  $p = 1/2$  Bernoulli trial with a biased coin by

- 1 Toss the coin twice
- 2 If you get two heads or two tails, repeat until you get HT or TH
- 3 If you get
  - ▶ HT call that a head
  - ▶ TH call that a tail

The two events HT and TH have equal probability, by construction.

# Source Coding and 20 Questions

Yet another way to think about coding

- 20 questions:
  - ▶ Want to guess a 'fact' — say an experiment's outcome
  - ▶ Only allowed Yes/No questions
  - ▶ Want to find the most efficient set of questions
- Obviously, Huffman code is optimal way of generating questions if we know the PMF

# Further reading I



Thomas M. Cover and Joy A. Thomas, *Elements of information theory*, John Wiley and Sons, 1991.