

Notational conventions
Class 13, 18.05, Spring 2014
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1 Learning Goals

1. Be able to work with the various notations and terms we use to describe probabilities and likelihood.

2 Notation and terminology for data and hypotheses

The problem of labeling data and hypotheses is a tricky one. When we started the course we talked about outcomes, e.g. heads or tails. Then when we introduced random variables we gave outcomes numerical values, e.g. 1 for heads and 0 for tails. This allowed us to do things like compute means and variances. We need to do something similar now. Recall our notational conventions:

- Events are labeled with capital letters, e.g. A, B, C .
- A random variable is capital X and takes values small x .
- The connection between values and events: ' $X = x$ ' is the event that X takes the value x .
- The probability of an event is capital $P(A)$.
- A discrete random variable has a probability mass function small $p(x)$ The connection between P and p is that $P(X = x) = p(x)$.
- A continuous random variable has a probability density function $f(x)$ The connection between P and f is that $P(a \leq X \leq b) = \int_a^b f(x) dx$.
- For a continuous random variable X the probability that X is in the infinitesimal interval $x \pm dx/2$ is $f(x) dx$.

In the context of Bayesian updating we will use similar conventions.

- We use capital letters, especially \mathcal{H} , to indicate a hypothesis, e.g. $\mathcal{H} =$ 'the coin is fair'
- We use lower case letters, especially θ , to indicate the hypothesized value of a model parameter, e.g. the probability the coin lands heads is $\theta = .5$
- We use upper case letters, especially \mathcal{D} , when talking about data as events. For example, $\mathcal{D} =$ 'the sequence of tosses was HTH.
- We use lower case letters, especially x , when talking about data as values. For example, the sequence of data was $x_1, x_2, x_3 = 1, 0, 1$.

- When the set of hypotheses is discrete we can use the probability of individual hypotheses, e.g. $p(\theta)$. When the set is continuous we need to use the probability for an infinitesimal range of hypotheses, e.g. $f(\theta) d\theta =$ probability of the range $\theta \pm d\theta/2$.

The following table summarizes this for discrete θ and continuous θ . In both cases we are assuming a discrete set of possible outcomes (data) x . Tomorrow we will deal with a continuous set of outcomes.

	hypothesis	prior	likelihood	unnormalized	
	\mathcal{H}	$P(\mathcal{H})$	$P(\mathcal{D} \mathcal{H})$	posterior	posterior
	\mathcal{H}	$P(\mathcal{H})$	$P(\mathcal{D} \mathcal{H})$	$P(\mathcal{D} \mathcal{H})P(\mathcal{H})$	$P(\mathcal{H} \mathcal{D})$
Discrete θ :	θ	$p(\theta)$	$p(x \theta)$	$p(x \theta)p(\theta)$	$p(\theta x)$
Continuous θ :	$\theta \pm d\theta/2$	$f(\theta) d\theta$	$p(x \theta)$	$p(x \theta)f(\theta) d\theta$	$f(\theta x) d\theta$

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