# **Prob/Stats Cheatsheet**

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ABSTRACT: Everything I know about prob/stats/maybe information theory too..

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#### 1 Conventions

#### **Math Notation**

#### 2 Distributions

### 2.1 Gaussians

1. To start with, memorize that

$$\int_{-\infty}^{\infty} dx \, e^{-x^2} = \sqrt{\pi} \tag{2.1}$$

2. Next, anything multiplying the  $x^2$  in the integrand is present in inverse under the square root.

$$\int_{-\infty}^{\infty} dx \, e^{-\text{stuff} \, x^2} = \sqrt{\frac{\pi}{\text{stuff}}}$$
 (2.2)

so, for example:

$$\int_{-\infty}^{\infty} dx \, e^{-\frac{1}{2}ax^2} = \sqrt{\frac{2\pi}{a}}$$
 (2.3)

#### 2.2 Bernoulli

For  $x \in \{0, 1\}$ , Bernoulli dist parametrized by  $\mu$ , with

$$p(x; \mu) = \mu^{x} (1 - \mu)^{1 - x}$$
(2.4)

#### 3 Prob and stats

#### 3.1 The Rules of Probability

• **Product Rule**: p(x, y) = p(x|y)p(y) = p(y|x)p(x)

• Sum Rule: 
$$p(x) = \sum_{y} p(x, y) = \sum_{y} p(x|y)p(y)$$

#### 3.2 Bayes' Rule

Using p(y|x)p(x) = p(x, y) = p(x|y)p(y), we have

$$p(y|x) = \frac{p(x|y)p(y)}{p(x)} = \frac{p(x|y)p(y)}{\sum_{y} p(x|y)p(y)}$$
 (3.1)

#### 3.3 Covariance

## 4 Information Theory

#### KL divergence:

$$KL[p(x)||q(x)] = \sum_{x_i} p(x_i) \log \left(\frac{p(x_i)}{q(x_i)}\right) = -\sum_{x_i} p(x_i) \log \left(\frac{q(x_i)}{p(x_i)}\right)$$

$$= -\sum_{x_i} p(x_i) \log q(x_i) + \sum_{x_i} p(x_i) \log p(x_i)$$

$$= H(p, q) - H(p)$$

$$(4.1)$$

where H(p, q) is the cross entropy, and H(p) is the entropy.

## 5 Bayesian

## 6 Optimal Stopping Theory