



Probability Distributions: Continuous

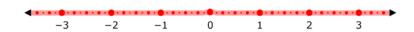
Introduction to Data Science Algorithms

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- Continious random variables
- Uniform
- Exponential
- Gamma
- Pareto
- Normal
- Multvariate normal used in machine learning
- · Dirichlet used in machine learning

Continuous random variables

- Today we will look at continuous random variables:
 - Real numbers: \mathbb{R} ; $(-\infty, \infty)$
 - Positive real numbers: \mathbb{R}^+ ; $(0, \infty)$
 - Real numbers between -1 and 1 (inclusive): [-1,1]
- The sample space of continuous random variables is uncountably infinite.



Continuous distributions

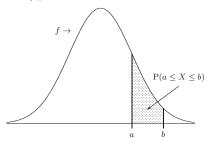
- Last time: a discrete distribution assigns a probability to every possible outcome in the sample space
- How do we define a continuous distribution?
- Suppose our sample space is all real numbers, ℝ.
 - What is the probability of P(X = 20.1626338)?
 - What is the probability of P(X = -1.5)?

Continuous distributions

- Last time: a discrete distribution assigns a probability to every possible outcome in the sample space
- How do we define a continuous distribution?
- Suppose our sample space is all real numbers, ℝ.
 - What is the probability of P(X = 20.1626338)?
 - What is the probability of P(X = -1.5)?
- The probability of any continuous event is always 0.
 - Huh?
 - There are infinitely many possible values a continuous variable could take. There is zero chance of picking any one exact value.
 - We need a slightly different definition of probability for continuous variables.

Probability density

- A probability density function (PDF, or simply density) is the continuous version of probability mass functions for discrete distributions.
- The probability at any point is zero, but the probably within a range is defined. For example, $\int_a^b f(x)$ is $P(a \le X \le b)$.



• Easier to express using the Cummulative Distribution Function (CDF): $P(a \le X \le b) = F(b) - F(a)$.

 While the probability for a specific value is 0 under a continuous distribution, we can still measure the probability that a value falls within an interval.

$$P(X \ge a) = \int_{x=a}^{\infty} f(x) = 1 - F(a)$$

$$P(X \le a) = \int_{x=-\infty}^{a} f(x) = F(a)$$

•
$$P(a \le X \le b) = \int_{x=a}^{b} f(x) = F(b) - F(a)$$

- This is analogous to the disjunction rule for discrete distributions.
 - For example if X is a die roll, then $P(X \ge 3) = P(X = 3) + P(X = 4) + P(X = 5) + P(X = 6)$
 - An integral is similar to a sum

Likelihood

- The likelihood function refers to the PMF (discrete) or PDF (continuous).
- For discrete distributions, the likelihood of x is P(X = x).
- For continuous distributions, the likelihood of x is the density f(x).
- We will often refer to likelihood rather than probability/mass/density so that the term applies to either scenario.
- The Cummualative Distribution Function $F_X(x) = P(X \le x)$ is valid for either discrete or continious distributions.

Uniform Distribution

- X ~ U(a, b): X is distributed as the Uniform distribution over a and b, for b > a.
- PDF is $f(x) = \frac{1}{b-a}$ for $a \le x \le b$.
- CDF $F(x) = \frac{x-a}{b-a}$ for $a \le x \le b$.
- Random number generators typically return samples U(0,1).
- These values are used to generate other random variates

Exponential distribution

- The exponential distribution is over positive real numbers (including zero), with the highest density at zero and decaying as x increases
- Sample space: [0, ∞)
- The probability density function is:

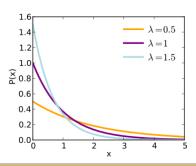
$$f(x) = \lambda e^{-\lambda x}$$

$$F(X) = 1 - e^{-\lambda x}$$

- Exponential is *memoryless*: P[T > s + t | T > s] = P[T > t]
- If interarrival time is $\sim E(\lambda)$ then number of arrivals is Poission distribution with parameter $1/\lambda$. Fundemental to queueing theory, less so for data science.

Exponential distribution

- A good model for:
 - The length of a phone call
 - Time between packets in the internet
 - The time between shooting stars during a meteor shower
 - The distance between cracks in a pipeline
- The parameter $\lambda > 0$ controls how quickly the density decays.



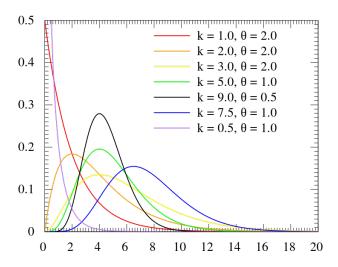
Gamma distribution

- The gamma distribution is a generalization of the exponential distribution (and others)
- Two parameters: shape k > 0, scale $\theta > 0$
- PDF:

$$f(x) = \frac{x^{k-1} \exp(-\frac{x}{\theta})}{\theta^k \Gamma(k)}$$

• Equivalent to exponential distribution when k = 1, $\theta = \frac{1}{\lambda}$

Gamma distribution



Pareto Distribution

- Pareto distribution: $X \sim f(\alpha)$ where $f(x) = \alpha/x^{\alpha+1}$
- Vilfredo Pareto noticed that income was distributed $\sim C/x^{\alpha}$ for some constants X and $\alpha > 0$.

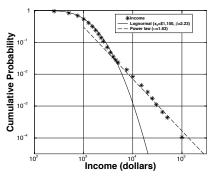


Fig. 1. The power law and lognormal fits to the 1935-36 U.S. income data. The solid line represents the lognormal fit with $x_0 = \$1$, 100 and $\beta = 2.23$. The straight dashed line represents the power law fit with $\alpha = 1.63$ (Badger 1980, with permission from Taylor & Francis Ltd.).