

# Quant Curriculum Week 2

## Probability Theory and Statistics

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March 2, 2023

### 1 Discrete and Continuous Distributions

#### 1.1 Motivation

We will begin by studying distribution functions widely used for quantitative modeling. Having a intuitive understanding of these functions and being able to discern the unique characteristics of a distribution is a valuable skill.

#### 1.2 Definitions

1. Random Variable: can be discrete or continuous
2. Cumulative distribution function:  $F(a) = P\{X \leq a\}$ ,  $\int_{-\infty}^a f(x) dx$
3. Probability mass function:  $p(x) = P\{X = x\}$
4. Probability distribution function:  $f(x) = \frac{d}{dx}F(x)$
5. Expected value:  $E[X] = \sum_{x:p(x)>0} xp(x)$ ,  $\int_{-\infty}^{\infty} xf(x) dx$
6. Variance:  $var(X) = E[(X - E[X])^2] = E[X^2] - (E[X])^2$
7. Bernoulli trial: random experiment with two possible outcomes (success/failure) of consistent probability

#### 1.3 Discrete / Continuous Random Variable Distributions

1. Uniform: probability distributions with equally likely outcomes
2. Binomial (Discrete): probability distributions arising from several Bernoulli trials
3. Normal (Gaussian): continuous probability distribution associated with the Central Limit Theorem (often used for natural and social sciences)
4. Poisson (discrete): probability distribution expressing the probability of a given number of events occurring in a fixed interval of time given a known constant mean rate and independence
5. Gamma: a family of distributions related to the beta, exponential, and chi-squared distributions
6. Geometric (discrete): probability distribution of the number of Bernoulli trials to get one success
7. Exponential: probability distribution of the time between events in a Poisson point process
8. Lognormal: random variables with the log being normally distributed

## 1.4 Moments of Distributions

Moments of random variables describe quantitatively the shape of their distribution. For second and higher moments, the central moment (about the mean) is typically used to provide more meaningful information about the distribution's shape. If  $F$  is a cumulative probability distribution function, the  $n$ -th moment of the distribution is given by the Riemann-Stieltjes Integral:

$$\mu'(n) = E[X^n] = \int_{-\infty}^{\infty} x^n dF(x)$$

Furthermore, the Moment Generating Function is an alternative way to describe the distribution of a random variable. The Moment Generating Function of random variable  $X$ ,  $M_X(t)$ , is given by:

$$M_X(t) = E[e^{tX}] = \int_{-\infty}^{\infty} e^{tx} f(x) dx$$

The Moment Generating Function can be used to find the moments of a distribution, as  $M_X^{(n)}$  expanded becomes the series:

$$M_X(t) = E[e^{tX}] = 1 + tE(X) + \frac{t^2 E(X^2)}{2!} + \frac{t^3 E(X^3)}{3!} + \dots$$

Therefore, differentiating  $M_X(t)$   $i$  times and setting  $t = 0$  will give the  $i$ th moment about the origin. A problem with moment-generating functions is that unlike characteristic functions (Fourier transform series), they may not exist for certain distributions, as the integral does not necessarily converge. However, it exists for almost all common distributions.

1. First Moment: Mean / Expected Value (raw moment)
2. Second Moment: Variance, deviations from the mean (central moment)
3. Third Moment: Skewness, a measure of the asymmetry of a distribution about the mean. Right skewed is defined as a positive skewness with the tail extending towards the right side.
4. Fourth Moment: Kurtosis, a measure of the thickness of tails. Excess kurtosis is kurtosis excess of the normal distribution. In finance, kurtosis is a measure of the extent of price fluctuations (how often prices move dramatically) and can be used to model risk in Value At Risk models.

**Normal Moments Example** What is the moment-generating function of a standard normal distribution? What are the distribution's first, second, third, and fourth moments?

**Properties of Poisson Processes** Buses at a station arrive according to a Poisson process with an average arrival time of 10 minutes (0.1 / min). If you arrive at a random time, what is your expected waiting time? On average, how many minutes ago did the last bus leave?

Poisson Process has occurs at constant mean rate  $\lambda$ , events can not occur at the same time, and events occur independently. The Poisson distribution gives the number of events in some time interval  $t$  given expected number of events:

$$P(X = \lambda t) = \frac{\lambda t e^{-\lambda t}}{k!}$$

Meanwhile, the exponential distribution gives the expected time between events, and it is a continuous and memory-less distribution:

$$P(t) = \lambda e^{-\lambda t}$$

## 2 Quick Probability Problems

**Problem 2.1 Tennis Match** Two tennis players are playing a 3 set tennis match. The match ends when a player wins two sets. The probability of a player winning each set is constant throughout the match. If you are given money to bet on whether the game finishes in 2 or 3 sets, which outcome should you bet on?

**Problem 2.2 Aces** You shuffle a deck of 52 cards and begin to draw cards one by one without replacement. How many cards do you expect to draw for you to draw your first ace?

### 2.1 \*\*\*\*\* Trading Firm Technical Interview Round \*\*\*\*\*

This is a technical interview question (45 minutes) that was asked by a trading firm in their first round this recruiting season. (To be explained)