

# Quant Interview Review Sheet

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## 1 Preface

test 1.3 This document is intended as a review of the basic topics and tricks needed for a quantitative trading interview. It should be concise enough to be read as a refresher in the 15 minutes before an interview. Alternatively, it may be used as an aid when taking assessments as a reference for commonly used formulas.

## 2 Probability

### 2.1 Combinatorics

#### Useful Combinatorics Identities

- **Pascal's Rule:**

$$\binom{n}{r} = \binom{n-1}{r} + \binom{n-1}{r-1}$$

- **Symmetry:**

$$\binom{n}{r} = \binom{n}{n-r}$$

- **Sum over all combinations:**

$$\sum_{r=0}^n \binom{n}{r} = 2^n$$

- **Binomial Theorem:**

$$(x+y)^n = \sum_{k=0}^n \binom{n}{k} x^{n-k} y^k$$

#### Permutations with Identical Items

**Formula:**

Number of distinct permutations of  $n$  items with repeats:  $\frac{n!}{n_1! n_2! \cdots n_k!}$

where  $n_1, n_2, \dots, n_k$  are the counts of identical items.

**Use Case:** Arranging letters in a word with repeated letters (e.g., **BANANA**).

#### Stars and Bars

**Problem Type:** Distributing  $n$  identical objects into  $k$  distinct bins

**Formula:**

$$\text{Number of ways} = \binom{n+k-1}{k-1} = \binom{n+k-1}{n}$$

**Equivalent Problem:** Number of non-negative integer solutions to:

$$x_1 + x_2 + \cdots + x_k = n$$

**Variant - With Restrictions:** If each bin must have at least  $m$  objects:

Place  $m$  in each bin first, then distribute remaining:  $\binom{(n-km)+k-1}{k-1}$

#### Inclusion-Exclusion Principle

**Two Sets:**

$$|A \cup B| = |A| + |B| - |A \cap B|$$

**Three Sets:**

$$\begin{aligned} |A \cup B \cup C| &= |A| + |B| + |C| \\ &\quad - |A \cap B| - |A \cap C| - |B \cap C| \\ &\quad + |A \cap B \cap C| \end{aligned}$$

**General Form for  $n$  Sets:**

$$\left| \bigcup_{i=1}^n A_i \right| = \sum_{k=1}^n (-1)^{k-1} \sum_{1 \leq i_1 < \cdots < i_k \leq n} |A_{i_1} \cap \cdots \cap A_{i_k}|$$

## 2.2 Bayes' Theorem and Conditional Probability

### Conditional Probability, Bayes, and Law of Total Probability

Consider events  $A_1, \dots, A_n$ , which form a partition of the sample space as well as event  $B$ . Then,

$$P(A_1 | B) = \frac{P(A_1 \cap B)}{P(B)} = \frac{P(B | A_1)P(A_1)}{P(B)} = \frac{P(B | A_1)P(A_1)}{\sum_{i=1}^n P(B \cap A_i)} = \frac{P(B | A_1)P(A_1)}{\sum_{i=1}^n P(B | A_i)P(A_i)}$$

### De Morgan's Laws

**Set Theory Version:**

$$(A \cup B)^c = A^c \cap B^c$$
$$(A \cap B)^c = A^c \cup B^c$$

## 2.3 Expectation, Variance, Covariance, and Correlation

### Law of Total Expectation and Variance

For two random variables  $X, Y$  defined on the same sample space,

$$\mathbb{E}[X] = \mathbb{E}[\mathbb{E}[X | Y]]$$

$$\text{discrete: } \mathbb{E}[X] = \sum_{i=1}^{\infty} P(Y = y_i) \mathbb{E}[X | Y = y_i]$$

$$\text{continuous: } \mathbb{E}[X] = \int_{\mathbb{R}} \mathbb{E}[X | Y = y] f_Y(y) dy$$

In variance,

$$\text{Var}(X) = \text{Var}(\mathbb{E}[X | Y]) + \mathbb{E}[\text{Var}(X | Y)]$$

where

$$\text{Var}(X | Y) = \mathbb{E}[(X - \mathbb{E}[X | Y])^2 | Y]$$

The Law of Total Expectation says that if we “average over all averages” of  $X$  obtained by some information about  $Y$ , we obtain the true average. Similarly, the Law of Total Variance says that the true variance comes from two sources: between samples (the first term) and within samples (the second term).

### Covariance and Correlation

$$\text{Cov}(X, Y) = \mathbb{E}[XY] - \mathbb{E}[X]\mathbb{E}[Y]$$

$$\text{Corr}(X, Y) = \frac{\text{Cov}(X, Y)}{\sigma_X \sigma_Y}$$

Covariance and correlation are measurements of linear association of  $X$  and  $Y$ . An example of uncorrelated but not independent random variables are  $Z$  and  $Z^2$ , where  $Z \sim N(0, 1)$ .

### Properties of Expectation, Variance, and Covariance

Let  $a, b, c, d$  be real constants and  $X$  and  $Y$  be random variables with finite mean and variance. Then all of the following hold:

1.  $\mathbb{E}[aX + bY + c] = a\mathbb{E}[X] + b\mathbb{E}[Y] + c$
2.  $\text{Var}(aX + b) = a^2 \text{Var}(X)$
3.  $\text{Cov}(aX + b, cY + d) = ac \text{Cov}(X, Y)$
4.  $\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y) + 2 \text{Cov}(X, Y)$
5. If  $X$  and  $Y$  are independent and have finite mean, then  $X$  and  $Y$  are uncorrelated.
6.  $\text{Corr}(aX + b, cY + d) = \text{sign}(ac) \text{Corr}(X, Y)$
7.  $\text{Cov}(X, X) = \text{Var}(X)$
8. The correlation and covariance matrices are both positive semidefinite. ( $\forall \mathbf{x} \in \mathbb{R}^n, \quad \mathbf{x}^\top A \mathbf{x} \geq 0$ .)
9.  $|\text{Corr}(X, Y)| \leq 1$

## 2.4 Order Statistics

## 2.5 Distributions

Name	Modeling Intuition	PMF/PDF	CDF	MGF	$\mu$	$\sigma^2$
Bernoulli	Toss a coin, 1 if heads, else 0, coin lands heads with probability $p$	$P(X = x) = \begin{cases} p & \text{if } x = 1 \\ 1 - p & \text{if } x = 0 \end{cases}$	$P(X \leq x) = \begin{cases} 0 & \text{if } x < 0 \\ 1 - p & \text{if } 0 \leq x < 1 \\ 1 & \text{if } x \geq 1 \end{cases}$	$pe^\theta + (1 - p)$	$p$	$p(1 - p)$
Binomial	Toss a coin $n$ times, probability of $x$ heads, coin lands heads with probability $p$	$P(X = x) = \binom{n}{x} p^x (1 - p)^{n-x}$	$P(X \leq x) = \sum_{k=0}^{\lfloor x \rfloor} \binom{n}{k} p^k (1 - p)^{n-k}$	$[pe^\theta + (1 - p)]^n$	$np$	$np(1 - p)$
Geometric	Probability of tossing coin $x$ times until first heads, coin lands heads with probability $p$	$P(X = x) = p(1 - p)^{x-1}$	$P(X \leq x) = 1 - (1 - p)^x$	$\frac{pe^\theta}{1 - (1 - p)e^\theta}$	$\frac{1}{p}$	$\frac{1 - p}{p^2}$
Poisson	Probability of $x$ occurrences within a fixed time interval or space; parameter $\lambda$ represents average number of occurrences	$P(X = x) = \frac{\lambda^x e^{-\lambda}}{x!}$	$P(X \leq x) = \sum_{k=0}^{\lfloor x \rfloor} \frac{\lambda^k e^{-\lambda}}{k!}$	$e^{\lambda(e^\theta - 1)}$	$\lambda$	$\lambda$
Exponential	Probability distribution of time between events in a Poisson process occurring with rate $\lambda$	$f(x) = \lambda e^{-\lambda x}$ for $x \geq 0$	$P(X \leq x) = 1 - e^{-\lambda x}$ for $x \geq 0$	$\frac{\lambda}{\lambda - \theta}$ for $\theta < \lambda$	$\frac{1}{\lambda}$	$\frac{1}{\lambda^2}$
Uniform	Uniform distribution on interval $[a, b]$	$f(x) = \frac{1}{b - a}$ for $x \in [a, b]$	$P(X \leq x) = \begin{cases} 0 & \text{if } x < a \\ \frac{x - a}{b - a} & \text{if } a \leq x \leq b \\ 1 & \text{if } x > b \end{cases}$	$\frac{e^{b\theta} - e^{a\theta}}{\theta(b - a)}$ for $\theta \neq 0$	$\frac{a + b}{2}$	$\frac{(b - a)^2}{12}$
Normal	Continuous distribution with bell curve shape, parameterized by mean $\mu$ and variance $\sigma^2$	$f(x) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left(-\frac{(x - \mu)^2}{2\sigma^2}\right)$	$P(X \leq x) = \Phi\left(\frac{x - \mu}{\sigma}\right)$	$e^{\mu\theta + \frac{1}{2}\sigma^2\theta^2}$	$\mu$	$\sigma^2$

Table 1: Key Distributions and Their Properties

### **3 Statistics**

#### **3.1 Normal Distribution and Related Properties**

#### **3.2 Linear Regression**

#### **3.3 Hypothesis Testing**

#### **3.4 Confidence Intervals**

#### **3.5 Maximum Likelihood Estimation (MLE)**

## 4 Stochastic Processes

### 4.1 Markov Chains

### 4.2 Martingales

#### Martingales, Submartingales, Supermartingales

Let  $(\mathcal{F}_n)$  be a filtration and  $(X_n)$  an adapted process.

$$\text{Martingale: } \mathbb{E}[X_{n+1} \mid \mathcal{F}_n] = X_n$$

$$\text{Submartingale: } \mathbb{E}[X_{n+1} \mid \mathcal{F}_n] \geq X_n$$

$$\text{Supermartingale: } \mathbb{E}[X_{n+1} \mid \mathcal{F}_n] \leq X_n$$

#### Optimal Stopping Times

A stopping time  $\tau$  with respect to  $(\mathcal{F}_n)$  is a random time such that:

$$\{\tau \leq n\} \in \mathcal{F}_n \quad \text{for all } n$$

The optimal stopping problem seeks  $\tau$  to maximize:

$$\mathbb{E}[X_\tau]$$

for a given process  $(X_n)$ . The optimal stopping theorem provides conditions under which:

$$\mathbb{E}[X_\tau] \leq \mathbb{E}[X_0]$$

for all stopping times  $\tau$ , if  $(X_n)$  is a supermartingale.

### 4.3 Random Walks

#### Birth and Death Chains

A birth and death chain is a Markov chain on  $\mathbb{N}$  with transitions:

$$P(i, i+1) = \lambda_i \quad (\text{birth}), \quad P(i, i-1) = \mu_i \quad (\text{death})$$

$$P(i, i) = 1 - \lambda_i - \mu_i$$

with  $\lambda_i, \mu_i \geq 0$  and  $\lambda_i + \mu_i \leq 1$ .

### 4.4 Dynamic Programming

## 5 Finance

### 5.1 Options

### 5.2 Derivatives

### 5.3 Portfolio Theory

### 5.4 Arbitrage and Pricing



## 6 Brainteasers

### 6.1 Mathematical Sequences

#### Common Sequences in Quantitative Reasoning

**Arithmetic Sequence:** constant difference

$$a_n = a + (n - 1)d \quad \text{e.g. } 2, 5, 8, 11, \dots \quad (d = 3)$$

**Geometric Sequence:** constant ratio

$$a_n = ar^{n-1} \quad \text{e.g. } 3, 6, 12, 24, \dots \quad (r = 2)$$

**Mixed Operations:**

- $n^2 + 1$ : 2, 5, 10, 17, 26, ...
- $\times 2 + 1$ : 1, 3, 7, 15, 31, ...
- $n(n + 1)$ : 2, 6, 12, 20, 30, ...

**Powers and Factorials:**

- Squares: 1, 4, 9, 16, 25, ...
- Cubes: 1, 8, 27, 64, ...
- Factorials: 1, 2, 6, 24, 120, ... ( $n!$ )

**Prime Numbers:**

$$2, 3, 5, 7, 11, 13, 17, 19, 23, 29, \dots$$

**Alternating Patterns:**

- Alternate +3 and -1: 1, 4, 3, 6, 5, 8, ...

**Fibonacci-Type Recurrences:**

- $f_n = f_{n-1} + f_{n-2}$ : 0, 1, 1, 2, 3, 5, 8, ...
- $f_n = 3f_{n-1} + f_{n-2}$ : 2, 5, 17, 56, 185, 611

## 7 Algorithms and Data Structures

## References

- [1] The quant interview cheat sheet. <https://www.quantguide.io/Cheat-Sheet.pdf>. Accessed: August 14, 2025.
- [2] Quant Insider. Probability & statistics cheatsheet (linkedin post). [https://www.linkedin.com/posts/quant-insider\\_probability-statistics-cheatsheet-activity-7133740570425835522-60QN/](https://www.linkedin.com/posts/quant-insider_probability-statistics-cheatsheet-activity-7133740570425835522-60QN/). Accessed: August 14, 2025.