Summary of the book¹

A First Course in Quantitative Finance

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Notation

The empty set.

Given a set S, $A \subset S$ denotes that A is a subset of S. Following the convention of the book, there's no notational difference between proper and improper subsets.

 $S^{\mathtt{C}}$ Given a set S, S^{\complement} denotes the complement of S.

Given a set S, #S denotes the cardinality (number of elements) of S. $\mathbb R$ denotes the real numbers. #S

 \mathbb{R}

Definition symbol.

Part I

Technical Basics

1 A Primer on Probability

1.1 Probability and Measure

D. 1: Sample space

A set $\Omega = \{\omega_1, \omega_2, \dots\}$ with elementary states $\omega_1, \omega_2, \dots$ which may or may not realize is called a sample space. It is the set of all possible outcomes of an experiment.

D. 2: Event

An event is a set of elementary states of the world, for each of which we can tell with certainty whether or not it has realized after the random experiment is over.

Any subset E of the sample space, $E \subset \Omega$, is known as an event. In other words, an event is a set consisting of possible outcomes of the experiment.

D. 3: Complement

Let U be the set of all elements under study (the "universe") and let $A \subset U$. Then A^{\complement} is called the complement of A. A^{\complement} is the set of of elements that are not in A.

$$A^{\tt C} = \{x \in U : x \not\in A\}.$$

D. 4: σ -algebra

A family \mathcal{F} of sets (events) A, A_1, A_2, \ldots is called a σ -algebra, if it satisfies the following conditions

- (i) \mathcal{F} is nonempty, i.e., $\mathcal{F} \neq \emptyset$,
- (ii) if $A \in \mathcal{F}$ then $A^{\complement} \in \mathcal{F}$,
- (iii) if $A_1, A_2, \ldots \in \mathcal{F}$ then $\bigcup_{n=1}^{\infty} A_n \in \mathcal{F}$.

T. 1: De Morgan

The De Morgan's rule:

$$\bigcap_{n=1}^{\infty} A_n = \left(\bigcup_{n=1}^{\infty} A_n^{\mathfrak{g}}\right)^{\mathfrak{g}}.$$
 (1)

From this, the subsequent two laws, known as the De Morgan's laws, can be derived¹.

$$\bigcup_{n=1}^{\infty} A_n^{\mathfrak{g}} = \left(\bigcap_{n=1}^{\infty} A_n\right)^{\mathfrak{g}} \tag{2}$$

$$\bigcap_{n=1}^{\infty} A_n^{\mathfrak{g}} = \left(\bigcup_{n=1}^{\infty} A_n\right)^{\mathfrak{g}} \tag{3}$$

D. 5: Measurable space

Given a sample space Ω and a σ -algebra \mathcal{F} , the pair (Ω, \mathcal{F}) is called a measurable space.

D. 6: Power set

The power set² of a set S, denoted as 2^S , is the set of all subsets of S, including \emptyset and S itself.

¹See section 2.1 for derivations.

²The power set is often (like in this book) denoted as 2^S . The reason for this is that a power set of S has $2^{\#S}$ elements (subsets of S). Intuitively, one can either include an element of S in a subset or not, i.e., for each element of S there are two choices, leading to $2^{\#S}$ possible subsets.

D. 7: Borel- σ -algebra on \mathbb{R}

The Borel- σ -algebra on \mathbb{R} , denoted as $\mathcal{B}(\mathbb{R})$, is the σ -algebra generated by all open sets (a,b), where $a,b\in\mathbb{R}$ and $a\leq b$.

D. 8: Generated σ -algebra

The σ -algebra generated³ by the event A is $\mathcal{F} = \{\emptyset, A, A^{\complement}, \Omega\}$, denoted as $\sigma(A)$.

D. 9: Measure

A function $\mu: \mathcal{F} \to \mathbb{R}^+_0$, with the properties

(i)
$$\mu(\emptyset) = 0$$
,

(ii)
$$\mu\left(igcup_{n=1}^{\infty}A_{n}
ight)=\sum_{n=1}^{\infty}\mu(A_{n}), \, \text{for } A_{1},A_{2},\ldots \in \mathcal{F} \text{ and } A_{i}\cap A_{j}=\emptyset \text{ for } i\neq j,$$

is called a measure on the measurable space (Ω, \mathcal{F}) .

D. 10: Measure space

Given a measureable space (Ω, \mathcal{F}) and a measure μ on (Ω, \mathcal{F}) , the triple $(\Omega, \mathcal{F}, \mu)$ is called a measure space.

D. 11: Probability space

A measure space $(\Omega, \mathcal{F}, \mu)$ where the measure satisfies $\mu(\Omega) = 1$ is called a probability space. The associated measure μ is then called probability and is abbreviated as P(A) for $A \in \mathcal{F}$. Therefore, the probability space triple is written as (Ω, \mathcal{F}, P) .

1.2 Filtrations and the Flow of Information

D. 12: Filtration

The ascending sequence of σ -algebras \mathcal{F}_t , with $\mathcal{F}_0 \subset \mathcal{F}_t \subset \mathcal{F}$, is called a filtration.

If a filtration is generated by successively observing the particular outcomes of a process (like a coin toss), it is called the natural filtration of that process.

1.3 Conditional Probability and Independence

L. 1

Given a probability space (Ω, \mathcal{F}, P) and an event $A \in \mathcal{F}$ with P(A) > 0. Now define

$$\mathcal{F}_A = \{ A \cap B : B \in \mathcal{F} \},\$$

the family of all intersections of A with every event in \mathcal{F} . Then \mathcal{F}_A is itself a σ -algebra on A and the pair (A, \mathcal{F}_A) is a measurable space.

D. 13: Conditional probability

Given a probability space (Ω, \mathcal{F}, P) and two events $A, B \in \mathcal{F}$. For P(A) > 0, the probability measure $P(B \mid A)$ is called the conditional probability of B given A, and is defined as

$$P(B \mid A) = \frac{P(B \cap A)}{P(A)}.$$

L. 2

Given a probability space (Ω, \mathcal{F}, P) and an event $A \in \mathcal{F}$ such that P(A) > 0. Then the triple $(A, \mathcal{F}_A, P(\cdot \mid A))$ forms a new probability space.

³It can be shown that $\sigma(A)$ is the smallest σ -algebra containing A.

T. 2: Bayes' rule

$$P(B \mid A) = \frac{P(A \mid B)P(B)}{P(A)} = \frac{P(A \mid B)P(B)}{P(A \mid B)P(B) + P(A \mid B^{\complement})P(B^{\complement})}$$

D. 14: Independence

Two events A and B are said to be independent, if

$$P(A \cap B) = P(A)P(B)$$

A direct consequence of indepence is that if events A and B are independent, then the conditional probability of A given B collapses to the unconditional one:

$$P(A \mid B) = \frac{P(A \cap B)}{P(B)} = \frac{P(A)P(B)}{P(B)} = P(A).$$

2 Derivations and Proofs

2.1 De Morgan's Laws

The first law (Equation 2) can be derived by taking the complement on both sides of Equation 1:

$$\left(\bigcap_{n=1}^{\infty}A_n\right)^{\mathtt{c}}=\left(\left(\bigcup_{n=1}^{\infty}A_n^{\mathtt{c}}\right)^{\mathtt{c}}\right)^{\mathtt{c}}=\bigcup_{n=1}^{\infty}A_n^{\mathtt{c}}.$$

The second law (Equation 3) can be found by replacing A_n with $B_n \triangleq A_n^{\text{C}}$:

$$\bigcap_{n=1}^{\infty} B_n^{\mathbf{C}} \triangleq \bigcap_{n=1}^{\infty} A_n \stackrel{1}{=} \left(\bigcup_{n=1}^{\infty} A_n^{\mathbf{C}}\right)^{\mathbf{C}} \triangleq \left(\bigcup_{n=1}^{\infty} B_n\right)^{\mathbf{C}}.$$

Bibliography

 $[{\it Maz}18] \quad {\it Thomas Mazzoni}. \ {\it A First Course in Quantitative Finance}. \ {\it Cambridge University Press, 2018. \ DOI: } 10.1017/9781108303606.$