# Financial Engineering Exam

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S20

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- 1. Lebesgue integration theory
- 2. L<sup>p</sup> spaces
- 3. Decomposition of measures
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- 6. Fourier transform

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# Lebesgue integration theory Monotone Convergence Theorem

### Theorem (Monotone Convergence Thereom)

If  $(f_n)$  is a monotone increasing sequence of functions in  $M^+(X,m)$  which converges to f, then

$$\int f \, d\mu = \lim \int f_n \, d\mu. \tag{1.1}$$

#### Proof of Monotone Convergence Theorem

The strategy of the proof is to first show that

$$\lim \int f_n \, d\mu \le \int f \, d\mu, \tag{1.2}$$

then afterwards to show that also

$$\lim \int f_n \, d\mu \ge \int f \, d\mu, \tag{1.3}$$

in order to conclude that

$$\lim \int f_n \, d\mu = \int f \, d\mu \tag{1.4}$$

# Lebesgue integration theory Proof of Monotone Convergence Theorem

According to Corollary 2.10

Corollary (2.10)

If  $(f_n)$  is a sequence in M(X, m) which converges to f on X, the f is in M(X, m).

the function f is measurable.

#### Proof of Monotone Convergence Theorem

From Lemma 4.5(1.)

#### Lemma

1. If f and g belong to  $M^+(X, m)$  and  $f \leq g$ , then

$$\int f \, d\mu \le \int g \, d\mu. \tag{1.5}$$

2. If f belongs to  $M^+(X, m)$ , if E, F belong to m, and if  $E \subseteq F$ , then

$$\int_{E} f \, d\mu \le \int_{F} f \, d\mu. \tag{1.6}$$

we have that

$$\int f_n d\mu \leq \int f_{n+1} d\mu \leq \int f d\mu, \quad \forall n \in \mathbb{N}.$$
 (1.7)

#### Proof of Monotone Convergence Theorem

Therefore we must also have that

$$\lim \int f_n \, d\mu \le \int f \, d\mu. \tag{1.8}$$

So this was the first step of our strategy, now we proceed to the second step.

#### Proof of Monotone Convergence Theorem

Let  $\alpha\in\mathbb{R}$  be such that  $0<\alpha<1$  and let  $\varphi$  be a simple measurable function such that  $0\leq\varphi\leq f$ . Let

$$A_n = \{x \in X : f_n(x) \ge \alpha \varphi(x)\}, \qquad (1.9)$$

such that

- 1.  $A_n \in m$
- 2.  $A_n \subseteq A_{n+1}$
- 3.  $X = \bigcup A_n$

#### Proof of Monotone Convergence Theorem

According to Lemma 4.5

#### Lemma

1. If f and g belong to  $M^+(X, m)$  and  $f \leq g$ , then

$$\int f \, d\mu \le \int g \, d\mu. \tag{1.10}$$

2. If f belongs to  $M^+(X, m)$ , if E, F belong to m, and if  $E \subseteq F$ , then

$$\int_{E} f \, d\mu \le \int_{F} f \, d\mu. \tag{1.11}$$

it must be that

$$\int_{A_n} \alpha \varphi \, d\mu \le \int_{A_n} f_n \, d\mu \le \int f_n \, d\mu. \tag{1.12}$$

Proof of Monotone Convergence Theorem
Since the sequence A is monotone increasing and has union X, it follows from Lemma 4.3(2.) and Lemma 3.4(1.),

# Lemma (4.3)

1. If  $\varphi$  and  $\psi$  are simple functions in  $M^+(X, m)$  and c > 0, then

$$\int carphi\, d\mu = c\intarphi\, d\mu, \ (1.13)$$
  $\int (arphi+\psi)\, d\mu = \intarphi\, d\mu + \int\psi\, d\mu. \ (1.14)$ 

2. If  $\lambda$  is defined for E in m by

$$\lambda(E) = \int \varphi \chi_E \, d\mu, \quad (1.15)$$

then  $\lambda$  is a measure on m.

### Lemma (3.4)

Let  $\mu$  be a measure defined on a  $\sigma$ -algebra m.

1. If  $(E_n)$  is an increasing sequence in m, then

$$\mu\left(\bigcup_{n=1}^{\infty} E_n\right) = \lim \mu\left(E_n\right). \tag{1.16}$$

2. If  $(F_n)$  is a decreasing sequence in m and if  $\mu(F_1) < +\infty$ , then

$$\mu\left(\bigcap_{n=1}^{\infty}F_{n}\right)=\lim\mu\left(F_{n}\right).$$

(1.17)

#### Proof of Monotone Convergence Theorem

that

$$\int \varphi \, d\mu = \lim \int_{A_n} \varphi \, d\mu. \tag{1.18}$$

Taking the limit as n tends to infinity in (1.12) therefore gives that

$$\alpha \int \varphi \, d\mu \le \lim \int f_n \, d\mu. \tag{1.19}$$

Since this holds for all 0 <  $\alpha$  < 1, by taking the limit as  $\alpha$  tends to 1 we obtain

$$\int \varphi \, d\mu \le \lim \int f_n \, d\mu. \tag{1.20}$$

#### Proof of Monotone Convergence Theorem

As  $\varphi$  is any simple function in  $M^+$  such that  $0 \le \varphi \le f$ , we can conclude that

$$\int f \, d\mu = \sup_{\varphi} \int \varphi \, d\mu \le \lim_{\varphi} \int f_n \, d\mu, \tag{1.21}$$

which concludes the proof.

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# L<sup>p</sup> spaces Minkowski's Inequality

# Theorem (Minkowski's Inequality)

If f and h belong to  $L_p$ ,  $p \ge 1$ , then f + h belongs to  $L_p$  and

$$||f + h||_{p} \le ||f||_{p} + ||h||_{p}.$$
 (2.1)

#### Proof of Minkowski's Inequality

- ▶ The case of p = 1 is equivalent to the triangular inequality and will not be considered here.
- ▶ We suppose p > 1.

#### Proof of Minkowski's Inequality

As both f and h are measurable functions, so is the sum

$$f+h, (2.2)$$

as the sum of measurable functions are also measurable.

#### Proof of Minkowski's Inequality

From Corollary 5.4, Theorem 5.5 and the fact that

$$|f+h|^p \le [2\sup\{|f|,|h|\}]^p \le 2^p\{|f|^p+|h|^p\},$$
 (2.3)

it follows that  $f + h \in L^p$ .

#### Corollary (5.4)

If f is measurable, g is integrable, and  $|f| \le |g|$ , then f is integrable, and

$$\int |f| d\mu \le \int |g| d\mu. \quad (2.4)$$

### Theorem (5.5)

A constant multiple  $\alpha f$  and a sum f+g of functions in L belongs to L and

$$\int \alpha f \, d\mu, \qquad (2.5)$$

$$\int (f+g) d\mu = \int f d\mu + \int g d\mu. \quad (2.6)$$

#### Proof of Minkowski's Inequality

Using the triangular inequality we can see that

$$|f+h|^p = |f+h||f+h|^{p-1} \le |f||f+h|^{p-1} + |h||f+h|^{p-1}.$$
(2.7)

Since  $f + h \in L^p$  it implies that  $|f + h|^p \in L^1$ .

Assuming that  $\frac{1}{p} + \frac{1}{q} = 1$ , it implies that

$$\frac{1}{p} + \frac{1}{q} = 1 \Leftrightarrow 1 + \frac{p}{q} = p \Leftrightarrow \frac{p}{q} = p - 1 \Leftrightarrow p = (p - 1)q, (2.8)$$

and thereby that  $|f + h|^{p-1} \in L^q$ .

#### Proof of Minkowski's Inequality

We can now apply Hölder's Inequality,

### Theorem (Hölder's Inequality)

Let 
$$f \in L_p$$
 and  $g \in L_q$  where  $p > 1$  and  $(1/p) + (1/q) = 1$ . Then  $fg \in L_1$  and  $\|fg\|_1 \le \|f\|_p \|g\|_q$ .

to infer that

$$\int |f||f+h|^{p-1} d\mu \le ||f||_p \left\{ \int |f+h|^{(p-1)q} d\mu \right\}^{1/q}$$

$$= ||f||_p ||f+h||_p^{p/q}.$$
(2.9)

Note that exactly the same can be said for the other term

$$\int |h||f+h|^{p-1} d\mu \le ||h||_p \left\{ \int |f+h|^{(p-1)q} d\mu \right\}^{1/q} \tag{2.11}$$

$$= \|h\|_{p} \|f + h\|_{p}^{p/q}. \tag{2.12}$$

#### Proof of Minkowski's Inequality

This tells us that

$$||f + h||_{p}^{p} \le ||f||_{p} ||f + h||_{p}^{p/q} + ||h||_{p} ||f + h||_{p}^{p/q}$$

$$= \{||f||_{p} + ||h||_{p}\} ||f + h||_{p}^{p/q}.$$
(2.13)

If we let  $A = ||f + h||_p = 0$ , then the result becomes trivial as a norm by definition is greater than or equal to zero.

#### Proof of Minkowski's Inequality

Suppose now that  $A \neq 0$  then we can divide (2.14) by  $A^{p/q}$ 

$$\frac{A^{p}}{A^{p/q}} \le \{\|f\|_{p} + \|h\|_{p}\} \frac{A^{p/q}}{A^{p/q}} \tag{2.15}$$

$$A^{p-p/q} \le ||f||_p + ||h||_p, \tag{2.16}$$

by noting that p - p/q = 1, we obtain

$$||f + h||_{p} \le ||f||_{p} + ||h||_{p},$$
 (2.17)

which concludes the proof.

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#### Definition of Absolute Continuity

#### Definition (Absolutely Continuous)

A measure  $\lambda$  on m is said to be absolutely continuous with respect to a measure  $\mu$  on m if  $E \in m$  and  $\mu(E) = 0$  imply that  $\lambda(E) = 0$ . In this case we write  $\lambda << \mu$ .

Note that  $\mu$  can send more sets to 0 than  $\lambda$ , but not the other way around.

# Decomposition of measures Radón-Nikodym Theorem

### Theorem (Radón-Nikodym Theorem)

Let  $\lambda$  and  $\mu$  be  $\sigma$ -finite measures defined on m and suppose that  $\lambda$  is absolutely continuous with respects to  $\mu$ . Then there exists a function f in  $M^+(X,m)$ , such that,

$$\lambda(E) = \int_{E} f d\mu, \quad E \in m. \tag{3.1}$$

Moreover, the function f is uniquely determined  $\mu$ -almost everywhere.

#### Definition of Singular Measure

### Definition (A Singular Measure)

Two measures  $\lambda, \mu$  on m are said to be mutually singular if there are disjoint sets A, B in m, such that,  $X = A \cup B$  and  $\lambda(A) = \mu(B) = 0$ . In this case we write  $\lambda \perp \mu$ .

# Decomposition of measures Lebesgue Decomposition Theorem

### Theorem (Lebesgue Decomposition Theorem)

Let  $\lambda$  and  $\mu$  be sigma-finite measures defined on a sigma-algebra m. Then there exists a measure  $\lambda_1$  which is singular with respect to  $\mu$  and a measure  $\lambda_2$  which is absolutely continuous with respect to  $\mu$  such that  $\lambda=\lambda_1+\lambda_2$ . Moreover, the measures  $\lambda_1$  and  $\lambda_2$  are unique.

The Theorem is a consequence of the Radon-Nikodým Theorem.

#### Proof of Lebesgue Decomposition Theorem

It can be shown that a measure  $\nu$  can be decomposed

$$\nu = \lambda + \mu \tag{3.2}$$

such that  $\nu$  is  $\sigma$ -finite,  $\lambda << \nu$  and  $\mu << \nu$ .

A consequence of the Radon-Nikodým Theorem is that

$$\exists f, g \in M^+(X, m), \tag{3.3}$$

where

$$\lambda(E) = \int_{E} f \, d\nu, \quad \mu(E) = \int_{E} g \, d\nu, \qquad \forall \, E \in m. \tag{3.4}$$

#### Proof of Lebesgue Decomposition Theorem

Define the following sets

$$A := \{x \in X \mid g(x) = 0\}, \quad B := \{x \in X \mid g(x) > 0\}.$$
 (3.5)

As  $g \in M^+$  we have that

$$A \cup B = X, \quad A \cap B = \emptyset.$$
 (3.6)

Define now  $\lambda_1, \lambda_2 : m \longrightarrow \overline{\mathbb{R}}$  by

$$\lambda_1(E) := \lambda(E \cap A), \tag{3.7}$$

$$\lambda_2(E) := \lambda(E \cap B). \tag{3.8}$$

#### Proof of Lebesgue Decomposition Theorem

What we now want is to prove that

- 1.  $\lambda_1 \perp \mu$ ,
- 2.  $\lambda_2 << \mu$ .

We start by proving  $\lambda_1 \perp \mu$ . From the definition of A we get

$$\mu(A) = \int_{\text{def. of } \mu} \int_{A} g \, d\nu = 0$$

$$\lambda_{1}(B) = \lambda(B \cap A) = \lambda(\emptyset) = 0$$

$$\lambda_{1} \perp \mu.$$

$$\lambda_{1}(B) = \lambda(B \cap A) = \lambda(\emptyset) = 0$$

$$\lambda_{1} \perp \mu.$$

$$\lambda_{1}(B) = \lambda(B \cap A) = \lambda(\emptyset) = 0$$

$$\lambda_{1} \perp \mu.$$

$$\lambda_{1}(B) = \lambda(B \cap A) = \lambda(\emptyset) = 0$$

$$\lambda_{1} \perp \mu.$$

$$\lambda_{2}(B) = \lambda(B \cap A) = \lambda(B \cap A)$$

$$\lambda_{3}(B) = \lambda(B \cap A) = \lambda(B \cap A)$$

$$\lambda_{4}(B) = \lambda(B \cap A) = \lambda(B \cap A)$$

$$\lambda_{5}(B) = \lambda(B \cap A) = \lambda(B \cap A)$$

$$\lambda_{6}(B) = \lambda(B \cap A) = \lambda(B \cap A)$$

$$\lambda_{1}(B) = \lambda(B \cap A) = \lambda(B \cap A)$$

$$\lambda_{1}(B) = \lambda(B \cap A) = \lambda(B \cap A)$$

$$\lambda_{2}(B) = \lambda(B \cap A) = \lambda(B \cap A)$$

$$\lambda_{3}(B) = \lambda(B \cap A) = \lambda(B \cap A)$$

$$\lambda_{4}(B) = \lambda(B \cap A) = \lambda(B \cap A)$$

$$\lambda_{5}(B) = \lambda(B \cap A) = \lambda(B \cap A)$$

$$\lambda_{6}(B) = \lambda(B \cap A) = \lambda(B \cap A)$$

$$\lambda_{6}(B) = \lambda(B \cap A) = \lambda(B \cap A)$$

$$\lambda_{7}(B) = \lambda(B \cap A)$$

$$\lambda_{8}(B) = \lambda(B \cap A)$$

$$\lambda$$

This proves the first point.

#### Proof of Lebesgue Decomposition Theorem

Now we want to show that  $\lambda_2 << \mu$ .

If  $E \in m$  such that  $\mu(E) = 0$  then

$$\int_{E} g \, d\nu = 0. \tag{3.10}$$

By Corollary 4.10,

### Corollary (4.10)

Suppose that f belongs to  $M^+$ . Then f(x)=0  $\mu$ -almost everywhere on X if and only if

$$\int f \, d\mu. \tag{3.11}$$

this means that g(x) = 0  $\nu$ -almost everywhere on E.

#### Proof of Lebesgue Decomposition Theorem

Recall that the terminology  $\mu$ -almost everywhere means that there exists a subset  $N \in m$  with  $\mu(N) = 0$  such that the statement holds on the complement of N. Thus we have that

$$B := \{x \in X \mid g(x) > 0\}$$

$$\exists N \text{ s.t. } \nu(N) = 0$$

$$g(x) = 0 \text{ on } E \setminus N$$

$$\Rightarrow E \cap B \subseteq N \text{ outside } N, g(x) = 0$$

$$(3.12)$$

$$\underset{\mathsf{Lemma }}{\Longrightarrow} 0 \le \nu(E \cap B) \le \nu(N) = 0 \Rightarrow \nu(E \cap B) = 0. \tag{3.13}$$

#### Proof of Lebesgue Decomposition Theorem

But remembering that  $\lambda << \nu$  this implies that

$$\lambda(E \cap B) = 0 \implies \lambda_2(E) = 0. \tag{3.14}$$

Thus we have shown that  $\lambda_2 << \mu$ .

#### Proof of Lebesgue Decomposition Theorem

To show that  $\lambda = \lambda_1 + \lambda_2$  remeber that they were constructed from the sets A and B. We have that

$$X = A \cup B$$
 and  $A \cap B = \emptyset$ . (3.15)

For every measurable set E we have that

$$E = (E \cap A) \cup (E \cap B)$$
 and  $(E \cap A) \cap (E \cap B) = \emptyset$ . (3.16)

Therefore we het that from the countable additivity of a measure that

$$\lambda(E) = \lambda(E \cap A) + \lambda(E \cap B) \tag{3.17}$$

$$= \lambda_1(E) + \lambda_2(E) \tag{3.18}$$

# Decomposition of measures Proof of Lebesgue Decomposition Theorem

Because this is true for every measurable set we can say that in general

$$\lambda = \lambda_1 + \lambda_2. \tag{3.19}$$

We have now shown that there existence part of the theorem. To finish the proof we would have to also verify that the  $\lambda_1$  and  $\lambda_2$ , derived here, are unique.

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#### Tonelli's Theorem

### Theorem (Tonelli's Theorem)

Let  $(X, m, \mu)$  and  $(Y, n, \nu)$  be a  $\sigma$ -finite measure space and let F be a nonnegative measureable function on  $Z = X \times Y$  to  $\overline{\mathbb{R}}$ . Then the functions defined on X and Y by

$$f(x) = \int_{Y} F_{x} d\nu, \quad g(y) = \int_{X} F^{y} d\mu,$$
 (4.1)

are measureable and

$$\int_{X} f \, d\mu = \int_{Z} F \, d\pi = \int_{Y} g \, d\nu. \tag{4.2}$$

In other symbols,

$$\int_{X} \left( \int_{Y} F \, d\nu \right) \, d\mu = \int_{Z} F \, d\pi = \int_{Y} \left( \int_{X} F \, d\mu \right) \, d\nu. \tag{4.3}$$

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#### Approximation by nice functions

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### Fourier transform