

# Financial Engineering

## Exam

Kasper Rosenkrands

Aalborg University

S20

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1. Lebesgue integration theory
2.  $L^p$  spaces
3. Decomposition of measures
4. Generation of measures and product measures
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6. Fourier transform

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## Monotone Convergence Theorem

### Theorem (Monotone Convergence Theorem)

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

$$\int f \, d\mu = \lim \int f_n \, d\mu. \quad (1.1)$$

## Proof of Monotone Convergence Theorem

The strategy of the proof is to first show that

$$\lim \int f_n d\mu \leq \int f d\mu, \quad (1.2)$$

then afterwards to show that also

$$\lim \int f_n d\mu \geq \int f d\mu, \quad (1.3)$$

in order to conclude that

$$\lim \int f_n d\mu = \int f d\mu \quad (1.4)$$

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.

## Lebesgue integration theory

### 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

$$\int f \, d\mu \leq \int g \, d\mu. \quad (1.5)$$

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

$$\int_E f \, d\mu \leq \int_F f \, d\mu. \quad (1.6)$$

we have that

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

Therefore we must also have that

$$\lim \int f_n d\mu \leq \int f d\mu. \quad (1.8)$$

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

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

such that

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

## Lebesgue integration theory

### 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

$$\int f \, d\mu \leq \int g \, d\mu. \quad (1.10)$$

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

$$\int_E f \, d\mu \leq \int_F f \, d\mu. \quad (1.11)$$

it must be that

$$\int_{A_n} \alpha \varphi \, d\mu \leq \int_{A_n} f_n \, d\mu \leq \int f_n \, d\mu. \quad (1.12)$$

## Lebesgue integration theory

### 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  $c \geq 0$ , then

$$\int c\varphi d\mu = c \int \varphi d\mu, \quad (1.13)$$

$$\int (\varphi + \psi) d\mu = \int \varphi d\mu + \int \psi d\mu. \quad (1.14)$$

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

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

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

$$\mu \left( \bigcap_{n=1}^{\infty} F_n \right) = \lim \mu(F_n). \quad (1.17)$$

## Proof of Monotone Convergence Theorem

that

$$\int \varphi d\mu = \lim \int_{A_n} \varphi d\mu. \quad (1.18)$$

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

$$\alpha \int \varphi d\mu \leq \lim \int f_n d\mu. \quad (1.19)$$

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

$$\int \varphi d\mu \leq \lim \int f_n d\mu. \quad (1.20)$$

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

$$\int f \, d\mu = \sup_{\varphi} \int \varphi \, d\mu \leq \lim \int f_n \, d\mu, \quad (1.21)$$

which concludes the proof. □

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## Minkowski's Inequality

### Theorem (Minkowski's Inequality)

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

$$\|f + h\|_p \leq \|f\|_p + \|h\|_p. \quad (2.1)$$

## 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

$$f + h, \tag{2.2}$$

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

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

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

### Theorem (5.5)

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

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

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

### Corollary (5.4)

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

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

### Proof of Minkowski's Inequality

Using the triangular inequality we can see that

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

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

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

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

## $L^p$ spaces

### 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  $\|fg\|_1 \leq \|f\|_p \|g\|_q$ .

to infer that

$$\int |f| |f + h|^{p-1} d\mu \leq \|f\|_p \left\{ \int |f + h|^{(p-1)q} d\mu \right\}^{1/q} \quad (2.9)$$

$$= \|f\|_p \|f + h\|_p^{p/q}. \quad (2.10)$$

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

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

$$= \|h\|_p \|f + h\|_p^{p/q}. \quad (2.12)$$

## Proof of Minkowski's Inequality

This tells us that

$$\|f + h\|_p^p \leq \|f\|_p \|f + h\|_p^{p/q} + \|h\|_p \|f + h\|_p^{p/q} \quad (2.13)$$

$$= \{\|f\|_p + \|h\|_p\} \|f + h\|_p^{p/q}. \quad (2.14)$$

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  $A^{p/q}$

$$\frac{A^p}{A^{p/q}} \leq \{\|f\|_p + \|h\|_p\} \frac{A^{p/q}}{A^{p/q}} \quad (2.15)$$

$$A^{p-p/q} \leq \|f\|_p + \|h\|_p, \quad (2.16)$$

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

$$\|f + h\|_p \leq \|f\|_p + \|h\|_p, \quad (2.17)$$

which concludes the proof.  $\square$

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## Decomposition of measures

### 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  $\lambda \ll \mu$ .

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,*

$$\lambda(E) = \int_E f d\mu, \quad E \in m. \quad (3.1)$$

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

## Decomposition of measures

### 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.

## Decomposition of measures

### 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,  $\lambda \ll \nu$  and  $\mu \ll \nu$ .

A consequence of the Radon-Nikodým Theorem is that

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

where

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

## Decomposition of measures

### Proof of Lebesgue Decomposition Theorem

Define the following sets

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

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

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

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

$$\lambda_1(E) := \lambda(E \cap A), \quad (3.7)$$

$$\lambda_2(E) := \lambda(E \cap B). \quad (3.8)$$

## Decomposition of measures

### Proof of Lebesgue Decomposition Theorem

What we now want is to prove that

1.  $\lambda_1 \perp \mu$ ,
2.  $\lambda_2 \ll \mu$ .

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

$$\left. \begin{array}{l} \mu(A) \underbrace{=}_{\text{def. of } \mu} \int_A g \, d\nu \underbrace{=}_{\text{def. of } A} 0 \\ \lambda_1(B) \underbrace{=}_{\text{def. of } \lambda_1} \lambda(B \cap A) \underbrace{=}_{\text{def. of } B} \lambda(\emptyset) = 0 \end{array} \right\} \lambda_1 \perp \mu. \quad (3.9)$$

This proves the first point.

## Decomposition of measures

### Proof of Lebesgue Decomposition Theorem

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

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

$$\int_E g \, d\nu = 0. \quad (3.10)$$

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*

$$\int f \, d\mu = 0. \quad (3.11)$$

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

## Decomposition of measures

### Proof of Lebesgue Decomposition Theorem

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

$$\left. \begin{array}{l} B := \{x \in X \mid g(x) > 0\} \\ \exists N \text{ s.t. } \nu(N) = 0 \\ g(x) = 0 \text{ on } E \setminus N \end{array} \right\} \Rightarrow E \cap B \underbrace{\subseteq}_{\text{outside } N, g(x) = 0} N \quad (3.12)$$

$$\underbrace{\Rightarrow}_{\text{Lemma 3.3}} 0 \leq \nu(E \cap B) \leq \nu(N) = 0 \Rightarrow \nu(E \cap B) = 0. \quad (3.13)$$



## Decomposition of measures

### Proof of Lebesgue Decomposition Theorem

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

$$\lambda(E \cap B) = 0 \implies \lambda_2(E) = 0. \quad (3.14)$$

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

## Decomposition of measures

### Proof of Lebesgue Decomposition Theorem

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

$$X = A \cup B \quad \text{and} \quad A \cap B = \emptyset. \quad (3.15)$$

For every measurable set  $E$  we have that

$$E = (E \cap A) \cup (E \cap B) \quad \text{and} \quad (E \cap A) \cap (E \cap B) = \emptyset. \quad (3.16)$$

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

$$\lambda(E) = \lambda(E \cap A) + \lambda(E \cap B) \quad (3.17)$$

$$= \lambda_1(E) + \lambda_2(E) \quad (3.18)$$

## Decomposition of measures

### Proof of Lebesgue Decomposition Theorem

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

$$\lambda = \lambda_1 + \lambda_2. \quad (3.19)$$

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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Approximation by nice functions

Proof of Lebesgue Decomposition Theorem



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Fourier transform

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