

## Lecture 4: Measures

(1)

We are going to define measures and probability measures in tandem so we can compare their properties.

Def: Let  $\mathcal{F}_0 \subset 2^{\Omega}$  be a field over  $\Omega$

$P$  is a Probability measure on  $(\Omega, \mathcal{F}_0)$  if

- $P: \mathcal{F}_0 \rightarrow [0, 1]$
- $P(\Omega) = 1$
- $P\left(\bigcup_{k=1}^{\infty} A_k\right) = \sum_{k=1}^{\infty} P(A_k)$   
when  $\bigcup_{k=1}^{\infty} A_k \in \mathcal{F}_0$  &  $A_k$ 's are disjoint

$\mu$  is a measure on  $(\Omega, \mathcal{F}_0)$  if

- $\mu: \mathcal{F}_0 \rightarrow [0, \infty]$
- $\mu(\emptyset) = 0$
- $\mu\left(\bigcup_{k=1}^{\infty} A_k\right) = \sum_{k=1}^{\infty} \mu(A_k)$   
when  $\bigcup_{k=1}^{\infty} A_k \in \mathcal{F}_0$  &  $A_k$ 's are disjoint

Note: these definitions are over fields and allow me to state extension thms easily. (i.e. if  $P$  is a prob measure on  $B_0^{(\text{ord})}$  then  $\exists$  an extended prob measure on  $\sigma(B_0^{(\text{ord})})$ ).

Note: Any prob measure is also a measure since

$$\begin{aligned} \mathcal{F} \text{ is not empty} \Rightarrow & \exists A \in \mathcal{F} \\ \Rightarrow & \Omega = A \cup A^c \in \mathcal{F} \\ \phi = & \Omega^c \in \mathcal{F} \\ \Rightarrow & 1 = P(\Omega) = P(\phi) + \sum_{k=1}^{\infty} P(A_k) \\ \Rightarrow & P(\phi) = 0 \end{aligned}$$

Def:  $(\Omega, \mathcal{F})$  is a measurable space if  $\mathcal{F}$  is a  $\sigma$ -field over  $\Omega$

Def:  $(\Omega, \mathcal{F}, P)$  is called a Probability Space:

$\sigma$ -field  $\uparrow$  Prob measure on  $\mathcal{F}$

$(\Omega, \mathcal{F}, \mu)$  is called a measure space measure on  $\mathcal{F}$

Def:

Let  $(\Omega, \mathcal{F}, \mu)$  be a measure space.

$\mu$  is finite if  $\mu(\Omega) < \infty$ .

$\mu$  is infinite if  $\mu(\Omega) = \infty$ .

$\mu$  is  $\sigma$ -finite over  $\mathcal{F}$  if

$\exists A_1, A_2, \dots \in \mathcal{F}$  s.t.

$$\Omega = \bigcup_{k=1}^{\infty} A_k \quad \&$$

$$\mu(A_k) < \infty$$

$\mu$  is  $\sigma$ -finite if  $\mu$  is  $\sigma$ -finite over  $\mathcal{F}$ .

Remark: Most of the measures people work with are  $\sigma$ -finite. In fact, we will show that all non-trivial  $\sigma$ -finite measures can be represented as integrals  $\mu(A) = \int_A dP$  where  $P$  is prob (The probabilist's world view of measure theory).

Remark:  $\exists$  FAP that are not measures.

e.g.  $\mathcal{F}_0 =$  finite & co-finite subsets of an infinite sample space.

$$P(A) = \begin{cases} 0 & \text{if } A \text{ is finite} \\ 1 & \text{if } A \text{ is co-finite} \end{cases}$$

Now  $P$  is a FAP,  $P$  is a measure if  $\Omega$  is uncountable but  $P$  is not a measure if  $\Omega$  is countable.

# Probability

v/s

# Measure

## Example

$$\mathcal{R} = \{1, 2, \dots, n\}$$

$$\mathcal{F} = 2^{\mathcal{R}}$$

$$P(A) = \sum_{w \in A} I_A(w) = \# \text{ of elements in } A$$

P is the uniform probability on  $\{1, 2, \dots, n\}$

## Basic Properties

Suppose P is a probability measure on a field  $\mathcal{F}_0$  over  $\mathcal{R}$

Then...

$$1. P(A^c) = 1 - P(A)$$

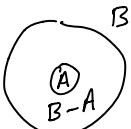
$$\begin{aligned} \text{since } P(\mathcal{R}) &= P(A \cup A^c) \\ &= P(A) + P(A^c) \end{aligned}$$

$$2. P(\emptyset) = 0$$

3. If  $A \subset B$  then

$$P(B - A) = P(B) - P(A).$$

$$\text{since } B = (B - A) \cup A$$



$$4. A \subset B \implies P(A) \leq P(B)$$

$$\begin{aligned} \text{by 3 since } 0 &\leq P(B - A) \\ &= P(B) - P(A) \end{aligned}$$

## Example

$\mathcal{R}$  is any set

$$\mathcal{F} = 2^{\mathcal{R}}$$

$$\mu(A) = \sum_{w \in A} I_A(w)$$

defined as  
the sup over  
finite sums.  
Can prove a version  
of tubini to prove  
the measure  
axioms hold

$\mu$  is called counting measure.

## Basic Properties

Suppose  $\mu$  is a measure on a field  $\mathcal{F}_0$  over  $\mathcal{R}$

Then...

$$1. \mu(A^c) = \mu(\mathcal{R}) - \mu(A) \text{ if } \mu(A) < \infty$$

$$\text{since } \mu(\mathcal{R}) = \mu(A) + \mu(A^c)$$

but can only move  $\mu(A)$  over if finite

$$2. \text{ SAME by definition}$$

$$3. \text{ If } A \subset B \text{ & } \mu(A) < \infty \text{ then}$$

$$\mu(B - A) = \mu(B) - \mu(A).$$

$$4. \text{ SAME, since we still have}$$

$$\mu(B) = \underbrace{\mu(B - A)}_{> 0} + \mu(A)$$

# Probability

vrs

# Measure

(24)

$$5. P(A \cup B) = P(A) + P(B) - P(A \cap B)$$

use  $\overset{A}{\text{---}} \cap \overset{B}{\text{---}} = A \cup (B - A \cap B)$

and 3.

$$6. P\left(\bigcup_{k=1}^n A_k\right) \leq \sum_{k=1}^n P(A_k)$$

can overlap

use 5 &  
induction.

$$7. \text{ If } A_n \uparrow A \in \mathcal{F}_0 \text{ then } P(A_n) \uparrow P(A)$$

Note: property 7 is called continuous  
from below

Proof:

$$P(A_n) = P\left(\bigcup_{k=1}^n A_k\right) \text{ since } A_1 \subset A_2 \subset \dots$$

$$= P\left(\bigcup_{k=1}^n \underbrace{[A_k - A_{k-1}]}_{\text{Same as } A_n - (A_1 \cup \dots \cup A_{k-1})} \right)$$

so disjoint

$$= \sum_{k=1}^n P(A_k - A_{k-1}), \quad A_0 := \emptyset$$

$$\uparrow \sum_{k=1}^{\infty} P(A_k - A_{k-1})$$

$$= P\left(\bigcup_{k=1}^{\infty} [A_k - A_{k-1}]\right)$$

$$= P(A)$$

since  $P$  is  
a Prop measure

$$\left. \begin{aligned} 5. \mu(A \cup B) &= \mu(A) + \mu(B - A \cap B) \\ &= \mu(A) + \mu(B) - \mu(A \cap B) \\ &\uparrow \\ &\text{if } \mu(A \cap B) < \infty \end{aligned} \right\}$$

6. SAME. For proof use  
the first part of 5 & induction

7. SAME, same proof too

disjointing  
technique

(5)

## Probability

vrs

## Measure

8. If  $A_n \downarrow A \in \mathcal{F}_0$  then  $P(A_n) \downarrow P(A)$ .  
 Called continuity from above

### Proof

First note that  $A_n \downarrow A \Leftrightarrow A_n^c \uparrow A^c$

By 7. we have

$$\begin{aligned} & P(A_n^c) \uparrow P(A^c) \\ & = 1 - P(A_n) \quad = 1 - P(A) \end{aligned}$$

$\therefore P(A_n) \downarrow P(A)$ .

9.  $P\left(\bigcup_{k=1}^{\infty} A_k\right) \leq \sum_{p=1}^{\infty} P(A_p)$ .  
 Called countable subadditivity

Proof:  $\bigcup_{p=1}^n A_p \uparrow \bigcup_{p=1}^{\infty} A_p$

$$\begin{aligned} \therefore P\left(\bigcup_{p=1}^{\infty} A_p\right) &= \lim_n P\left(\bigcup_{p=1}^n A_p\right) \\ &= \lim_n \sum_{p=1}^n P(A_p - A_1 \cup \dots \cup A_{p-1}) \\ &\leq \lim_n \sum_{p=1}^n P(A_p) \end{aligned}$$

8. If  $A_n \downarrow A \in \mathcal{F}_0$  and  $\mu(A_m) < \infty$  for some  $m$  then  $\mu(A_n) \downarrow \mu(A)$ .

Note: the extra condition is necessary.  
 e.g.  $\Omega = \mathbb{Z}, \mathcal{F}_0 = 2^{\mathbb{Z}}, \mu = \text{counting measure}$   
 $A_n = \{n, n+1, \dots\} \downarrow A = \emptyset$  but

$$\underbrace{\mu(A_n)}_{\infty} \downarrow \underbrace{\mu(A)}_{=0}$$

### Proof:

Let  $n \geq m$  & note  $A_m \supset A_n \supset A$ .  
 $\therefore \mu(A) \leq \mu(A_n) \leq \mu(A_m) < \infty$

$$\text{Define } P(B) := \frac{\mu(B \cap A_m)}{\mu(A_m)}$$

$P$  is a prob measure on  $(\Omega, \mathcal{F}_0)$ .

$\therefore P(A_n) \downarrow P(A)$  by 8 for Probs

$$\therefore \frac{\mu(A_n \cap A_m)}{\mu(A_m)} \downarrow \frac{\mu(A \cap A_m)}{\mu(A_m)}$$

$\therefore \mu(A_n) \downarrow \mu(A)$  since  $A_n \cap A_m = A_n$   
 $\& A \cap A_m = A$ .

9. SAME. SAME proof too.

# Probability

vrs

# Measure

## Thm (uniqueness)

Let  $P$  &  $Q$  be two prob measures on  $(\mathbb{R}, \sigma(\mathcal{P}))$ . If

- (i)  $P = Q$  on  $\mathcal{P}$
- (ii)  $P$  is a  $\pi$ -sys

Then  $P = Q$  on  $\sigma(\mathcal{P})$ .

Proof: Use "good sets".

$$\text{Let } \mathcal{Y} = \{A \in \sigma(\mathcal{P}) : P(A) = Q(A)\}$$

We want to show  $\sigma(\mathcal{P}) \subseteq \mathcal{Y}$ .

By Dynkin's  $\pi$ - $\lambda$  extension of good sets

we need  $P \subseteq \mathcal{Y}$   
 ↓  
 $\pi$ -system       $\lambda$ -sys.

Clearly  $P \subseteq \mathcal{Y}$  &  $P$  is a  $\pi$ -sys.

To show  $\mathcal{Y}$  is a  $\lambda$ -sys just note:

•  $\mathbb{R} \in \mathcal{Y}$ , since  $P(\mathbb{R}) = Q(\mathbb{R}) = 1$

•  $A \in \mathcal{Y} \Rightarrow P(A) = Q(A)$

$$\Rightarrow P(A^c) = Q(A^c)$$

$$\Rightarrow A^c \in \mathcal{Y}$$

•  $\underbrace{A_1, A_2, \dots}_{\text{disjoint}} \in \mathcal{Y} \Rightarrow P\left(\bigcup_{k=1}^{\omega} A_k\right) = \sum_{k=1}^{\omega} P(A_k)$

$$= \sum_{k=1}^{\omega} Q(A_k)$$

$$= Q\left(\bigcup_{k=1}^{\omega} A_k\right)$$

$$\Rightarrow \bigcup_{k=1}^{\omega} A_k \in \mathcal{Y}.$$

$$Q \in \mathcal{D}.$$

## Thm (uniqueness)

Let  $\mu$  &  $\nu$  be two measures on  $(\mathbb{R}, \sigma(\mathcal{P}))$ . If

- (i)  $\mu = \nu$  on  $\mathcal{P}$

- (ii)  $\mathcal{P}$  is a  $\pi$ -sys

- (iii)  $\mu, \nu$  are  $\sigma$ -finite over  $\mathcal{P}$

Then  $\mu = \nu$  on  $\sigma(\mathcal{P})$ .

Note: The extra condition is necessary.

For example,  $\mathcal{P} = \{[0, x] : 0 < x \leq 1\}$ ,

$$\mu(A) = \sum_{w \in \mathcal{P}} I_A(w) \quad \text{i.e. counting measure}$$

$$\nu(A) = \sum_{w \in \mathcal{P}} I_{A \cap \text{rational}}(w)$$

$\mu \neq \nu$  both agree on  $\mathcal{P}$  but not on  $\sigma(\mathcal{P})$

since  $\mu(\text{irrationals}) = \infty$

$$\nu(\text{irrationals}) = 0$$

Notice this set is in  $\mathcal{B}^{[0,1]}$

Proof:

Let  $c_1, c_2, \dots \in \mathcal{P}$  which cover  $\mathbb{R}$   
 and  $\mu(c_k), \nu(c_k) < \infty$ . Define

$$P_k(A) = \begin{cases} \mu(A \cap c_k) / \mu(c_k) & \text{if } \mu(c_k) > 0 \\ 0 & \text{o.w.} \end{cases}$$

$$Q_k(A) = \begin{cases} \nu(A \cap c_k) / \nu(c_k) & \text{if } \nu(c_k) > 0 \\ 0 & \text{o.w.} \end{cases}$$

case 1:  $\mu(c_k) = 0$

$$\therefore \nu(c_k) = 0$$

$$\therefore P_k = Q_k \text{ on } \sigma(\mathcal{P})$$

Case 2:  $\mu(C_k) > 0$

$\therefore v(C_k) > 0$  & both  $P_k, Q_k$  are prob measures.

Since  $A \in \mathcal{P} \Rightarrow A \cap C_k \in \mathcal{P}$  we have  $P_k = Q_k$  on  $\mathcal{P}$ .

Both  $P_k$  &  $Q_k$  are prob measures.

$\therefore P_k = Q_k$  on  $\sigma(\mathcal{P})$

Now stitch  $P_k$  &  $Q_k$  together

$$\begin{aligned} \mathcal{D} = \bigcup_{k=1}^{\infty} C_k &= \bigcup_{k=1}^{\infty} C_k - C_1 \cup \dots \cup C_{k-1} \\ &= \bigcup_{k=1}^{\infty} C_k \cap C_1^c \cap \dots \cap C_{k-1}^c \\ &\quad \text{disjoint cover} \\ &\quad \dots \text{but not in } \mathcal{P}. \end{aligned}$$

$$\begin{aligned} \therefore \mu(A) &= \mu\left(\bigcup_{k=1}^{\infty} A \cap C_k \cap C_1^c \cap \dots \cap C_{k-1}^c\right) \\ &= \sum_{k=1}^{\infty} \mu(A \cap C_k \cap C_1^c \cap \dots \cap C_{k-1}^c) \\ &= \sum_{k=1}^{\infty} \mu[C_k] P_k(A \cap C_1^c \cap \dots \cap C_{k-1}^c) \\ &= \sum_{k=1}^{\infty} v[C_k] Q_k(A \cap C_1^c \cap \dots \cap C_{k-1}^c) \\ &= v(A) \end{aligned}$$

QED

(7)

Continuity is equivalent to countable additivity for  $P$

(8)

The next thm is only needed for probability & is useful for showing a  $P$  is measurable

Thm:

If  $P$  is a finitely additive prob on a field  $\mathcal{F}_0 \subset 2^{\mathbb{N}}$  then the following are equivalent

- (i)  $P$  is a prob measure
- (ii)  $P$  is continuous from below
- (iii)  $P$  is continuous from above
- (iv)  $\forall A_1, A_2, \dots \in \mathcal{F}_0$  s.t.  $A_n \downarrow \emptyset$  one has  $\underbrace{P(A_n)}_{\text{continuous from above at } \emptyset} \downarrow 0$

Proof:

We already have

$$(i) \Rightarrow (ii) \Rightarrow (iii) \Rightarrow (iv).$$

Show (iv)  $\xrightarrow{(a)} (ii) \xrightarrow{(b)} (i)$

For (a) suppose (iv) & let  $\overbrace{A_n \uparrow A}$  all in  $\mathcal{F}_0$

$$\therefore A_n^c \downarrow A^c$$

$$\therefore A_n^c - A^c \downarrow \emptyset$$

$$\therefore P(A_n^c - A^c) \downarrow 0 \text{ by (iv)}$$

$$\therefore P(A_n^c) - P(A^c) \downarrow 0, \text{ since } A_n^c \subset A^c \text{ so that } A^c \subset A_n^c$$

$$\therefore -P(A_n) + P(A) \downarrow 0$$

$$\therefore P(A_n) \uparrow P(A)$$

∴ (ii) holds

Note: these properties follow by FAP

For (b) suppose (ii) holds (9)

Let  $\underbrace{A_1, A_2, \dots}_{\text{disjoint}} \in \mathcal{F}_0$  &  $\bigcup_{k=1}^n A_k \in \mathcal{F}_0$

$$\begin{aligned}\therefore P\left(\bigcup_{k=1}^n A_k\right) &= \lim_n P\left(\bigcup_{k=1}^n A_k\right) \\ &= \lim_n \sum_{k=1}^n P(A_k) \\ &= \sum_{k=1}^{\infty} P(A_k)\end{aligned}\quad \text{QED}$$

### Null & Negligible sets

Definitions: Let  $(\mathcal{R}, \mathcal{F}, \mu)$  be a measure space.

- $A$  is  $\mu$ -null iff  $A \in \mathcal{F}$  &  $\mu(A) = 0$ .
- $A$  is  $\mu$ -negligible iff  $\exists$  a  $\mu$ -null cover of  $A$ .
- $(\mathcal{R}, \mathcal{F}, \mu)$  is complete iff all  $\mu$ -neg subsets of  $\mathcal{R}$  are in  $\mathcal{F}$
- The completion of  $\mathcal{F}$  w.r.t  $\mu$  is  $\bar{\mathcal{F}} = \sigma(\mathcal{F}, \eta_\mu)$

where  $\eta_\mu := \{N \subset \mathcal{R} : N \text{ is } \mu\text{-neg}\}$ .

Once we prove the existence of Lebesgue measure  $\mathcal{L}$  on  $\mathbb{R}$  (10)

$$\mathcal{B}^\mathbb{R} := \sigma(B^\mathbb{R}, \eta_\mathcal{L})$$

= Lebesgue measurable sets of  $\mathbb{R}$

where  $B^\mathbb{R}$  = Borel measurable sets of  $\mathbb{R}$ .

The following thm shows how to extend  $(\mathcal{R}, \mathcal{F}, \mu)$  to  $(\mathcal{R}, \bar{\mathcal{F}})$ .

claim: If  $(\mathcal{R}, \mathcal{F}, \mu)$  is a measure space then  $\sigma(\mathcal{F}, \eta_\mu) = \{F \cup N : F \in \mathcal{F}, N \in \eta_\mu\}$

Proof: "RHS."

The inclusion " $\supset$ " is trivial.

For " $\subset$ " use good sets.

- Clearly  $\mathcal{F}, \eta_\mu \subset \text{RHS}$  since  $\emptyset \in \mathcal{F}$  &  $\emptyset \in \eta_\mu$
- $\emptyset \in \text{RHS}$ .

- RHS is closed under countable unions since  $\bigcup_k (F_k \cup N_k) = (\bigcup_k F_k) \cup (\bigcup_k N_k)$

- $A \in \text{RHS} \Rightarrow A = F \cup N$  for a  $\mu$ -null  $C \supset A$

$$A^c = F^c \cap N^c$$

$$= (F^c \cap N^c) \cap (C \cup C^c)$$

$$= (F^c \cap N^c \cap C^c) \cup (F^c \cap N^c \cap C)$$

since  $C^c \subset N^c$   
 $\therefore C^c \cap N^c = C^c$

$\hookrightarrow = C^c$   $\in \eta_\mu$  since covered by  $C$

$$A^c \in \text{RHS} \subset \mathcal{F}$$

QED.

Now we can define  $\bar{\mu}$  on  $\bar{F}$  (11)

by  $\bar{\mu}(F \cup N) = \mu(F)$

$\uparrow$        $\downarrow$   
 $\in F$        $\in \mathcal{N}_\mu$

Claim:

- (i)  $\bar{\mu}$  is well defined
- (ii)  $\bar{\mu}$  is a measure on  $(\mathcal{R}, \bar{F})$  extending  $\mu$  on  $(\mathcal{R}, F)$ .
- (iii)  $\bar{\mu}$  is unique
- (iv)  $(\mathcal{R}, \bar{F}, \bar{\mu})$  is complete.

Proof:

For (i): Suppose  $F_1 \cup N_1 = F_2 \cup N_2$

$$\therefore F_1 - F_2 \subset N_2 \quad \& \quad F_2 - F_1 \subset N_1$$

For suppose  $w \in F_1 - F_2$

$$\therefore w \in F_1 \quad \& \quad w \notin F_2$$

$$\therefore w \in \underbrace{F_1 \cup N_1}_{= F_2 \cup N_2} \quad \& \quad w \notin F_2$$

$$\therefore w \in N_2$$

$$\therefore \mu(F_1 - F_2) = \mu(F_2 - F_1) = 0$$

$$\therefore \mu(F_1) \leq \mu(F_1 \cup F_2)$$

$$= \mu(F_2 \cup (F_1 - F_2))$$

$$= \mu(F_2) + \underbrace{\mu(F_1 - F_2)}$$

$$\& \mu(F_2) \leq \mu(F_1) \text{ similarly}$$

$$\therefore \mu(F_1) = \mu(F_2) \quad (12)$$

$$\therefore \bar{\mu}(F_1 \cup N_1) = \bar{\mu}(F_2 \cup N_2).$$

For (ii): only need to show

$$\bar{\mu}\left(\bigcup_k (F_k \cup N_k)\right) = \bar{\mu}\left(\left[\bigcup_k F_k\right] \cup \left[\bigcup_k N_k\right]\right)$$

disjoint over  $k$        $\in F$        $\in \mathcal{N}_\mu$

$$:= \mu\left(\bigcup_k F_k\right)$$

$$\begin{aligned} &\stackrel{\substack{F_k \cup N_k \text{ disjoint} \\ (\text{implies } F_k \text{'s disjoint})}}{=} \sum_k \mu(F_k) \\ &= \sum_k \bar{\mu}(F_k \cup N_k) \end{aligned}$$

For (iii):

Let  $\nu$  be a measure on  $(\mathcal{R}, \bar{F})$

s.t.  $\nu = \mu$  on  $F$ .

$$\therefore \nu = \mu \text{ on } \mathcal{N}_\mu$$

$$\therefore \nu(F \cup N) = \nu(F) = \mu(F) = \bar{\mu}(F \cup N)$$

}

$\geq$  by "increasing"

$\leq$  by sub-additivity

$$\therefore \nu = \mu \text{ on } \bar{F}.$$

For (iv) let  $\bar{N} \subset \mathcal{R}$  be  $\bar{\mu}$ -neg

$$\therefore \bar{N} \subset F \cup N \subset F \cup A$$

$$\begin{cases} \text{for } F \in F \text{ & } N \in \mathcal{N}_\mu \\ \text{s.t. } 0 = \bar{\mu}(F \cup N) = \mu(F) \end{cases} \quad \begin{cases} \text{for } A \in F \text{ s.t.} \\ \mu(A) = 0 \end{cases}$$

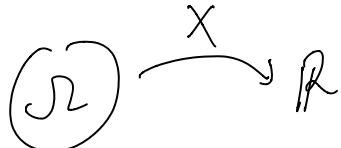
$\therefore \bar{N}$  is  $\mu$ -neg so  $\bar{F}$  contains all  $\bar{\mu}$ -neg sets

QED

What are we going to use measures for

(13)

Random variables will be maps



where  $(\Omega, \mathcal{F}, P)$  is a prob space &  $X$  is  $\mathcal{F}$ -measurable.

$$E[X] = \int_{\Omega} X(\omega) dP(\omega)$$

↑  
Integration w.r.t.  
measure  $P$

For two measures  $P$  &  $\Omega$  on  $(\Omega, \mathcal{F})$  the Radon-Nikodym

derivative  $\frac{dP|X^{-1}}{d\Omega|X^{-1}}$  will be

The likelihood ratio of  $P$  to  $\Omega$  for  $X$ .

Radon-Nikodym derivatives will also give use conditional expected values, etc.

Why does probability seem so rich a subject when its essentially just measure theory under the constraint  $\mu(\Omega) = 1$

(14)

I think the answer is that once  $\mu(\Omega) = 1$  you can understand  $\Omega$  as modeling a random draw  $w \in \Omega$ . So the extra assumption essentially gives an isomorphism from the abstract  $(\Omega, \mathcal{F}, \mu)$  to a physical random procedure.

So a solution can be made on the abstract structure  $(\Omega, \mathcal{F}, \mu)$  or the physical random draws it models.