

My name is  
stochastic calculus  
but everybody calls me  
calculus

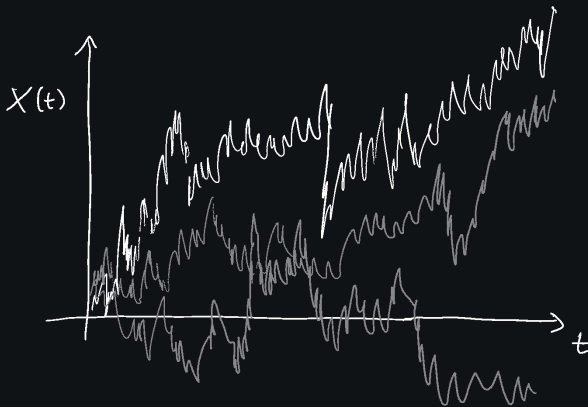
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March 11th, 2021

(Day 434 of the COVID Era)

# Idea

"random variables"



"almost everywhere"

$\bigcup_n \bigcap_m A_m$  events as propositions

# Preliminaries

Let's fix a probability space  $W = (W, \mathcal{F}, \mathbb{P})$ . Recall that

$W$  : **Set**

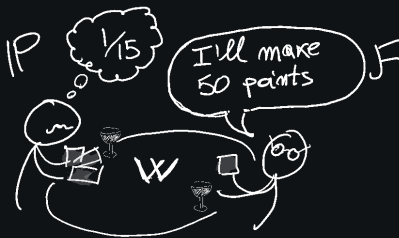
Space of 'outcomes' of a stochastic experiment

$\mathcal{F}$  :  $\sigma$ -field

Space of 'events', i.e.  $\sigma$ -algebra of subsets of  $W$

$\mathbb{P} : \mathcal{F} \rightarrow \mathbb{R}^+$

'Probability measure', i.e. an assignment of beliefs to events



# Preliminaries

We call

$$\ker \mathbb{P} = \{A \in \mathcal{F} \mid \mathbb{P}(A) = 0\}$$

the  $\sigma$ -ideal of null sets.

Let  $V = (V, \mathcal{G}, \mathbb{Q})$ .

measurable maps

$$\mathbf{Msbl}(W, V) = \{f : W \rightarrow V \mid f^{-1}\mathcal{G} \subseteq \mathcal{F}\}$$

null-reflecting maps

$$\mathbf{Msbl}_0(W, V) = \{f \in \mathbf{Msbl}(W, V) \mid f^{-1} \ker \mathbb{Q} \subseteq \ker \mathbb{P}\}$$



# Preliminaries

## Definition

An  $E$ -valued **stochastic process** on  $W$  is a collection of measurable maps

$$X_t : W \rightarrow E, \quad t \in I$$

where  $I$  is a poset of **times** and  $E$  is a measurable space of **values**.

## Example

Toss a coin and bet 1\$ on heads:  $W = \{H, T\}^{\mathbb{N}}$ ,  $E = \mathbb{R}$ ,  $I = \mathbb{N}$ ,

$$X_n(\omega) = \sum_{i=0}^n (\delta_H^{\omega_i}(\omega) - \delta_T^{\omega_i}(\omega))$$

## Example

You have a financial portfolio, your profit are a stochastic process  $X_t$  which is a function of another stochastic process  $M_t$ , the market.



# Preliminaries

A central concept in stochastic calculus is that of filtration:

## Definition

A **filtration** on  $W$  is a sequence of  $\sigma$ -fields  $\{\mathcal{F}_t\}_{t \in I}$  such that  $\mathcal{F}_t \subseteq \mathcal{F}_s$  when  $t \leq s$ . One sets  $\mathcal{F}_\infty := \mathcal{F}$ .

It gives a 'causal structure' to the observed events.

## Definition

An **adapted process** is a process  $\{X_t\}_{t \in I}$  such that

for each  $t \in I$ ,  $X_t$  is  $\mathcal{F}_t$ -measurable.

i.e. the value of  $X$  at  $t$  only depends on info available until that point.



# Preliminaries

**Itô calculus:** a popular way to do stochastic calculus

1. We can **integrate** wrt stochastic processes:

$$\int_0^t X_t \, dB_t = \lim_n \sum_i X_{t_i}^{(n)} (B_{t_{i+1}} - B_{t_i})$$

2. We can discuss **differential problems**:

$$dX_t = \sigma_t X_t dB_t + \mu_t X_t dt$$

Here  $B_t$  is a **Brownian motion**  $\rightsquigarrow$  archetypal diffusion process

It's not exactly like elementary calculus though, e.g.

$$df(B_t) = f'(B_t) dB_t + \frac{1}{2} f''(B_t) dt$$

because  $B_t$  has 'unbounded variation'

# Idea

Original goal:

‘elementary’  
stochastic calculus  
topos

It turned out to be much harder than I thought  
and probably not that simple

*...still, plenty of interesting stuff along the way!*

**Warning:** work still in progress!



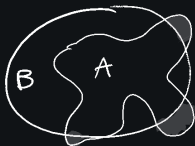


# Essential algebra

Define the preorder  $(\mathcal{F}, \leq)$ :

$$A \leq B \quad \text{iff} \quad \mathbb{P}(A \Rightarrow B) = 1$$

i.e. if  $A$  is essentially contained in  $B$ :



$$\mathbb{P}(\emptyset) = 0$$

The **essential algebra** of  $\mathcal{W}$  is the *posetal reflection* of  $(\mathcal{F}, \leq)$ :

$$\mathbb{F} := (\mathcal{F} / \ker \mathbb{P}, \leq)$$

**Terminology:** informally, we use 'essential' to say 'up to a null set'.

# Tripes approach

## Theorem

$\mathbb{F}$  is a **complete** Boolean algebra.

## Proof.

$\mathbb{F}$  is  $\sigma$ -complete and every essential partition of  $W$  can be refined to a countable essential partition (CCC). □

In [Sco67], Dana Scott forces CH using a Boolean-valued model whose algebra of truth values is an essential algebra.

Therefore in [Cap20], we call

$$\mathbf{Scott}[W] := \mathbb{F}^{(-)} : \mathbf{Set}^{\mathrm{op}} \rightarrow \mathbf{Cat}$$

the **Scott tripes** associated to  $W$ .

# Scott tripos

This tripos captures ‘logic valued in  $W$ ’:

1. A predicate on  $X : \mathbf{Set}$  is a map  $\varphi : X \rightarrow \mathbb{F}$  ( *$W$ -predicates*).
2.  $\wedge, \vee, \Rightarrow$  are pointwise essential intersection/union/implication.
3.  $\text{true} = [W] = \textit{almost everywhere}$ ,  
 $\text{false} = [\emptyset] = \textit{almost nowhere}$ .
4. Quantifiers:

$$\forall x:X \varphi(x) := \bigwedge_{x \in X} \varphi(x), \quad \exists x:X \varphi(x) := \bigvee_{x \in X} \varphi(x)$$

5. Power objects:

$$(\pi(X), \in_X) = (\mathbb{F}^X, \text{ev} : \mathbb{F}^{\mathbb{F}^X} \times \mathbb{F}^X \rightarrow \mathbb{F}).$$



# Scott tripos

The tripos should be be 'hybridized' with quantitative notions from probability theory:

1. Given a proposition  $\varphi$ , we can talk about its probability  $\mathbb{P}(\varphi)$ .
2. Given a family of propositions  $\{\varphi_i\}_{i \in I}$ , we can talk about their independence.

## Open question

**Can this structure be captured by the tripos?**

I have some ideas about this... reach out if you want to talk about it.

# Tripes to topos

Since **Scott** $[W]$  is localic, its associated topos can be presented as

$$\mathrm{Sh} \mathbb{F} \simeq \mathrm{Sh}_{\mathrm{a.e.}} \mathcal{F} \hookrightarrow \mathrm{Sh} \mathcal{F}$$

where ‘a.e.’ is the almost-everywhere topology on  $\mathcal{F}$ :

$$\{A_i\}_{i \in I} \text{ covers } A \quad \underline{\text{iff}} \quad \{A_i\}_{i \in I} \text{ ‘essentially covers’ } A$$

## Definition

$$\mathrm{Sh}_{\mathrm{a.e.}} W := \mathrm{Sh} \mathbb{F} \simeq \mathrm{Sh}_{\mathrm{a.e.}} \mathcal{F} \simeq \mathbf{Pitts}[\mathbf{Scott}[W]]$$

It’s the topos of ‘random sets and random functions’ wrt to  $W$ .

It’s a classical universe: we have both LEM and AC.



# Geometry

We can look at this topos geometrically:

## algebraic geometry

base scheme  $S$

regular functions

structure sheaf  $\mathcal{O}_S$

commutative rings

Grothendieck topos  $\mathrm{Sh} S$

## probability theory

base probability space  $W$

essentially bounded r.v.

'structure sheaf'  $L^\infty(-, \mathbb{R})$

commutative Von Neumann algebras  
(kind of)

Kolmogorov topos  $\mathrm{Sh}_{\mathrm{a.e.}} W$



# Geometry

The rabbit hole runs deep...

Gelfand duality

Localizable measurable spaces

Riesz duality

Structured topoi

$\sigma$ -locales



# Geometry

There is a **measurability structure**

$$\mathcal{O}_W : \mathbf{Msbl} \rightarrow \mathbf{Sh}_{\text{a.e.}} W$$

given by **essential functors of points**

$$\mathcal{O}_W(E) = \mathbf{Msbl}(-, E) / \underset{\text{a.e.}}{=}$$

and any null-reflecting map  $f : W \rightarrow V$  induces a morphism

$$\begin{array}{ccc}
 \mathbf{Sh}_{\text{a.e.}} W & \begin{array}{c} \xleftarrow{f^*} \\ \xrightarrow{\perp} \\ \xrightarrow{f_*} \end{array} & \mathbf{Sh}_{\text{a.e.}} V \\
 \nwarrow \mathcal{O}_W & \swarrow \text{ } & \nearrow \mathcal{O}_V \\
 & \mathbf{Msbl} & 
 \end{array}$$

$$f^* \mathcal{O}_V \implies \mathcal{O}_W \quad \text{or, equivalently} \quad \mathcal{O}_V \implies f_* \mathcal{O}_W$$



# Internal notions

We can start looking for the objects of probability theory inside  $\text{Sh}_{\text{a.e.}} W$  (with the help of [Jac06]).

# Random variables

The internal real numbers are given by

$$\mathcal{R}_W = \mathcal{O}_W(\mathbb{R})$$

Hence this confirms ‘random variables’ are indeed ‘variables’ in a ‘random’ setting.

Then we can prove things like

## Theorem

*A sequence of random variables  $\{X_n\}_{n \in \mathbb{N}}$  converges almost surely iff it converges in the usual sense as a sequence of internal real numbers.*

Also,  $L^p_{\text{loc}}$  spaces become subsets of the real numbers.

# Integration

There's a sheaf  $\mathcal{M}$  of measures on  $W$ , which is definable internally

$$\mathcal{M} = \{\mu \in \text{a.e.}(P\mathbb{Q}^+) \mid \mu \text{ is an 'additive' upper cut}\}$$

Thus integration is straightforward: first, on simple rational functions:

$$\begin{aligned} \int -d- : \mathcal{Q}^+ \times \mathcal{M} &\longrightarrow \mathcal{M} \\ (q, \mu) &\longmapsto q\mu \end{aligned}$$

then we extend by continuity:

$$\begin{aligned} \int -d- : \mathcal{R}^+ \times \mathcal{M} &\longrightarrow \mathcal{M} \\ (X, \mu) &\longmapsto \underbrace{\bigvee_{q \leq X} q\mu}_{\int -X d\mu} \end{aligned}$$

In particular, we defined  $\mathbb{E}[X]$ .

# Radon–Nikodym

Now recall

## Theorem (Radon–Nikodym)

*Let  $\nu$  be a measure on  $\mathcal{W}$  such that  $\mu(A) = 0$  implies  $\nu(A) = 0$ . Then there exists a positive measurable random variable  $\frac{d\nu}{d\mu}$  such that*

$$\int_{-} \frac{d\nu}{d\mu} d\mu = \nu(-)$$

It says every absolutely continuous measure wrt  $\mu$  has a  $\mu$ -density.

# Radon–Nikodym

We can prove the theorem internally ([Jac06]). Fix  $\mu : \mathcal{M}$ :

1.  $\mu$ -integration in an arrow  $\int -d\mu : \mathcal{R}^+ \rightarrow \mathcal{M}$ .
2. Its image  $\mathcal{M}_{\int \mu}$  is contained in  $\mathcal{M}^{\ll \mu}$ , the object of absolutely continuous measures with respect to  $\mu$ .
3. Then we want to show  $\mathcal{M}_{\int \mu} = \mathcal{M}^{\ll \mu}$  so that  $\frac{d-}{d\mu}$  is the splitting of the image of  $\int -d\mu$ :

$$\frac{d-}{d\mu} : \mathcal{M}^{\ll \mu} \longrightarrow \mathcal{R}^+$$

To do so, one fixes  $\nu : \mathcal{M}^{\ll \mu}$  and defines

$$\frac{d\nu}{d\mu} = \bigvee \{q \in \mathcal{Q}^+ \mid q\mu \leq \nu\}$$

up to technicalities, this concludes the proof!

# Martingales

For stochastic calculus, the most useful application of Radon–Nikodym is given by the definition of **conditional expectations**:

## Definition

Given a filtration  $\{F_t\}_{t \in I}$ , the conditional expectation of  $X$  at  $t$  is defined as

$$\mathbb{E}[X|\mathcal{F}_t] = \frac{d(\int X \, d\mathbb{P})|_{\mathcal{F}_t}}{d\mathbb{P}|_{\mathcal{F}_t}}$$

## Definition

A **martingale** is an adapted stochastic process such that

$$\mathbb{E}[X_{t+s}|\mathcal{F}_t] = X_t$$

It's a process not expected to change, e.g. the coin tosses bet,  $B_t$ .

# Internalizing martingales

To define a martingale internally we need

1. An internal definition of 'stochastic process'
2. An internal definition of 'adaptedness'  $\rightsquigarrow$  **filtrations**
3. An internal definition of 'conditional expectation at  $t$ '.

# Stochastic processes

## Theorem

An  $E$ -valued stochastic process  $\{X_t\}_{t \in I}$  on  $W$  corresponds to a map

$$X : \Delta I \rightarrow \mathcal{O}_W(E)$$

in  $\text{Sh}_{a.e.} W$ .

If  $X$  is **measurable** as a map  $W \times I \rightarrow E$ , then this map can be lifted to

$$X : \mathcal{O}_W(I) \rightarrow \mathcal{O}_W(E).$$

By extension, any map  $X : I \rightarrow E$  in  $\text{Sh}_{a.e.} W$  is a ‘stochastic process’.



# Stochastic processes

Two facts:

## Theorem

Two stochastic processes  $X, Y : \Delta I \rightarrow \mathcal{O}_W(E)$  are equal in the logic of  $\text{Sh}_{a.e.} W$  iff they are **indistinguishable**, i.e.

$$\mathbb{P}(\forall t \in I, X_t = Y_t) = 1$$

Therefore we can state & prove theorems like:

## Theorem

A stochastic process is almost surely continuous iff it's continuous from the internal point of view.

# Filtrations

We need a way to express '*measurability at  $t$* ' for elements in  $\mathcal{O}_W(E)$ .

A filtration induces a chain of null-reflecting maps (each carried by  $1_W$ )

$$W_\infty \longrightarrow \cdots \longrightarrow W_t \longrightarrow \cdots \longrightarrow W_0$$

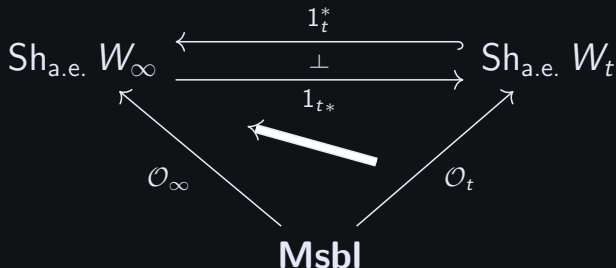
where  $W_t = (W, \mathcal{F}_t, \mathbb{P}|_{\mathcal{F}_t})$ .

which in turn becomes a chain of (surjections of) Kolmogorov topoi:

$$\mathrm{Sh}_{\mathrm{a.e.}} W_\infty \begin{array}{c} \xleftarrow{\perp} \\ \xrightarrow{\quad} \end{array} \cdots \begin{array}{c} \xleftarrow{\perp} \\ \xrightarrow{\quad} \end{array} \mathrm{Sh}_{\mathrm{a.e.}} W_t \begin{array}{c} \xleftarrow{\perp} \\ \xrightarrow{\quad} \end{array} \cdots \begin{array}{c} \xleftarrow{\perp} \\ \xrightarrow{\quad} \end{array} \mathrm{Sh}_{\mathrm{a.e.}} W_0$$

# Filtrations

Remember that null-reflecting maps induce maps



Hence for every  $t \in I$ ,  $E: \mathbf{Msbl}$ , we have

$$1_t^* \mathcal{O}_t(E) =: \mathcal{O}_t^*(E) \longrightarrow \mathcal{O}_\infty(E)$$

# Filtrations

The inclusion  $\mathbb{F}_t \hookrightarrow \mathbb{F}_\infty$  has both adjoints, in particular

$$\begin{aligned} 1_t^{-1} \dashv \Diamond_t : \mathbb{F}_\infty &\longrightarrow \mathbb{F}_t \\ A &\longmapsto \bigwedge \{B \in \mathcal{F}_t \mid A \leq B\} \end{aligned}$$

We can interpret it as a '*possibly at t*' modal operator.

Thus for each  $A \in \mathbb{F}_\infty$ :

$$\mathcal{O}_t^*(E)(A) := \mathcal{O}_t(E)(\Diamond_t A) = \mathbf{Msbl}(\Diamond_t A, E) / \underset{\text{a.e.}}{=}$$

and

$$\begin{aligned} \mathcal{O}_t^*(E)(A) &\longrightarrow \mathcal{O}_\infty(E)(A) \\ \Diamond_t A \xrightarrow{f} E &\longmapsto A \hookrightarrow \Diamond_t A \xrightarrow{f} E \end{aligned}$$

Notice:  $\Diamond_t A \subseteq W_t$ , while  $A \subseteq W_\infty$

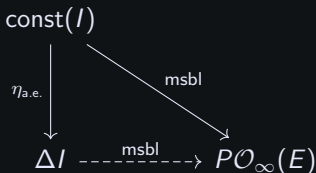
$\rightsquigarrow \mathcal{O}_t^*(E)$  contains  $\mathcal{F}_t$ -measurable elements.

# Filtrations

For each  $t \in I$ , let  $\text{msbl}(t)$  be the image of  $\mathcal{O}_t^*(E)$  in  $\mathcal{O}_\infty(E)$ .

Sheafification extends this definition: given  $\tau: \Delta I$ ,

$$W \models e \in \text{msbl}(\tau) \quad \text{iff} \quad \text{for all } s \in I, \{\tau = s\} \models e \in \text{msbl}(s)$$



We can extend the definition of  $\text{msbl}$  to  $\mathcal{O}_\infty(I)$ :

$$W \models \forall \tau: \mathcal{O}_\infty(I), \text{msbl}(\tau) \Leftrightarrow \forall \sigma: \Delta I, \sigma \leq \tau \Rightarrow \text{msbl}(\sigma)$$

This recovers  $\mathcal{F}_\tau$ -measurability for a **random time**  $\tau: W \rightarrow I$ .

# Filtrations

## Definition

We call  $X : \Delta I \rightarrow \mathcal{O}_\infty(E)$  **adapted** iff

$$W \models \forall t : \Delta I, X(t) \in \text{msbl}(t)$$

## Theorem

*External adaptedness coincides with internal adaptedness*

## Open question

Is there a better **internal/synthetic** notion of adaptedness?

**Alternative construction** suggested by Morgan Rogers (priv. comm.):

1. The object  $\mathcal{F}_\bullet = \{\mathcal{F}_t\}_{t \in I}$  is naturally in  $\text{Psh}(I, \leq)$
2. Adapted processes are maps of internal sheaves over  $\mathcal{F}_\bullet$ !

# Conditional expectation

Restrictions of measures is given by the canonical inclusion

$$1_{t*}\mathcal{M}_\infty \xhookrightarrow{-|_t} \mathcal{M}_t$$

Then we define

$$\mathbb{E}[-|\mathcal{F}_t]_t = 1_{t*}\mathcal{R}_\infty^+ \xrightarrow{(\int -d\mathbb{P})|_t} \mathcal{M}^{\ll \mathbb{P}}|_t \xrightarrow{\frac{d}{d\mathbb{P}}|_t} \mathcal{R}_t^+$$

Finally:

$$\mathbb{E}[-|\mathcal{F}_t]_\infty := \mathcal{R}_\infty^+ \xrightarrow{\text{ext. by } 0} 1_t^* 1_{t*}\mathcal{R}_\infty^+ \xrightarrow{1_t^* \mathbb{E}[X|\mathcal{F}_t]_t} 1_t^* \mathcal{R}_t^+ \xrightarrow{A \mapsto \Diamond_t A} \mathcal{R}_\infty^+$$

If we unpack the definitions, this is what's happening (hand-wavingly):

$$X \mapsto \bigvee \{q_t \in 1_t^* \mathcal{Q}_t^+ \mid q_t \leq (\int X \, d\mathbb{P})/\mathbb{P}\}$$

best approximation at  $t$  of  $X$ 's average.

# Martingales, finally

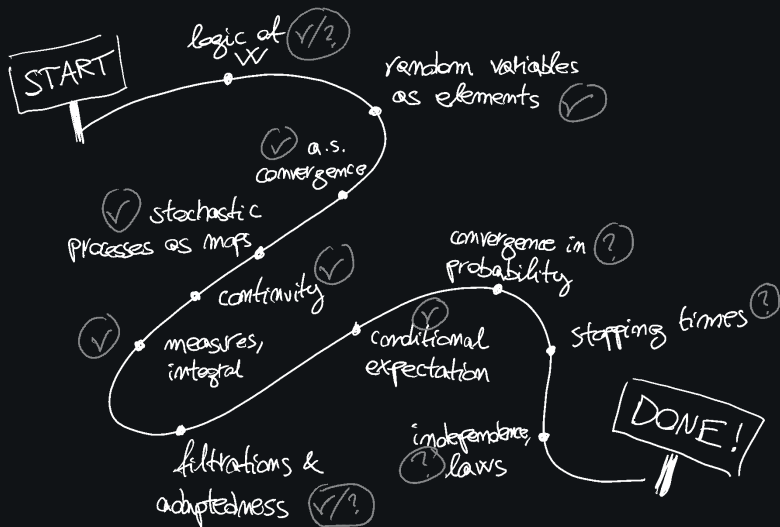
## Definition

A map  $X : \Delta I \rightarrow \mathcal{O}_\infty(E)$  is a **martingale** if it is adapted and

$$W \models \forall t : \Delta I \ \forall s \geq t, \ \mathbb{E}[X(t+s) | \mathcal{F}_t]_\infty = X(t),$$



# Roadmap



**Thanks for your attention!**

Questions?

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