

# Foundations for type-driven probabilistic modelling

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Computational golden era of:

logic & type rich  
computation

Statistical  
computation

# Computational golden era of:

logic & type rich  
computation

Expressive type systems:

Haskell, OCaml, Idris

Mechanised mathematics:

Agda, Coq, Isabelle/HOL, Lean

Verification:

SMT-powered, realistic  
systems

Statistical  
computation

generative modelling  
+

efficient inference:

Monte-Carlo simulation  
or gradient-based  
optimisation

"AI"

Computational golden era of:

logic type rich  
computation

Statistical  
computation

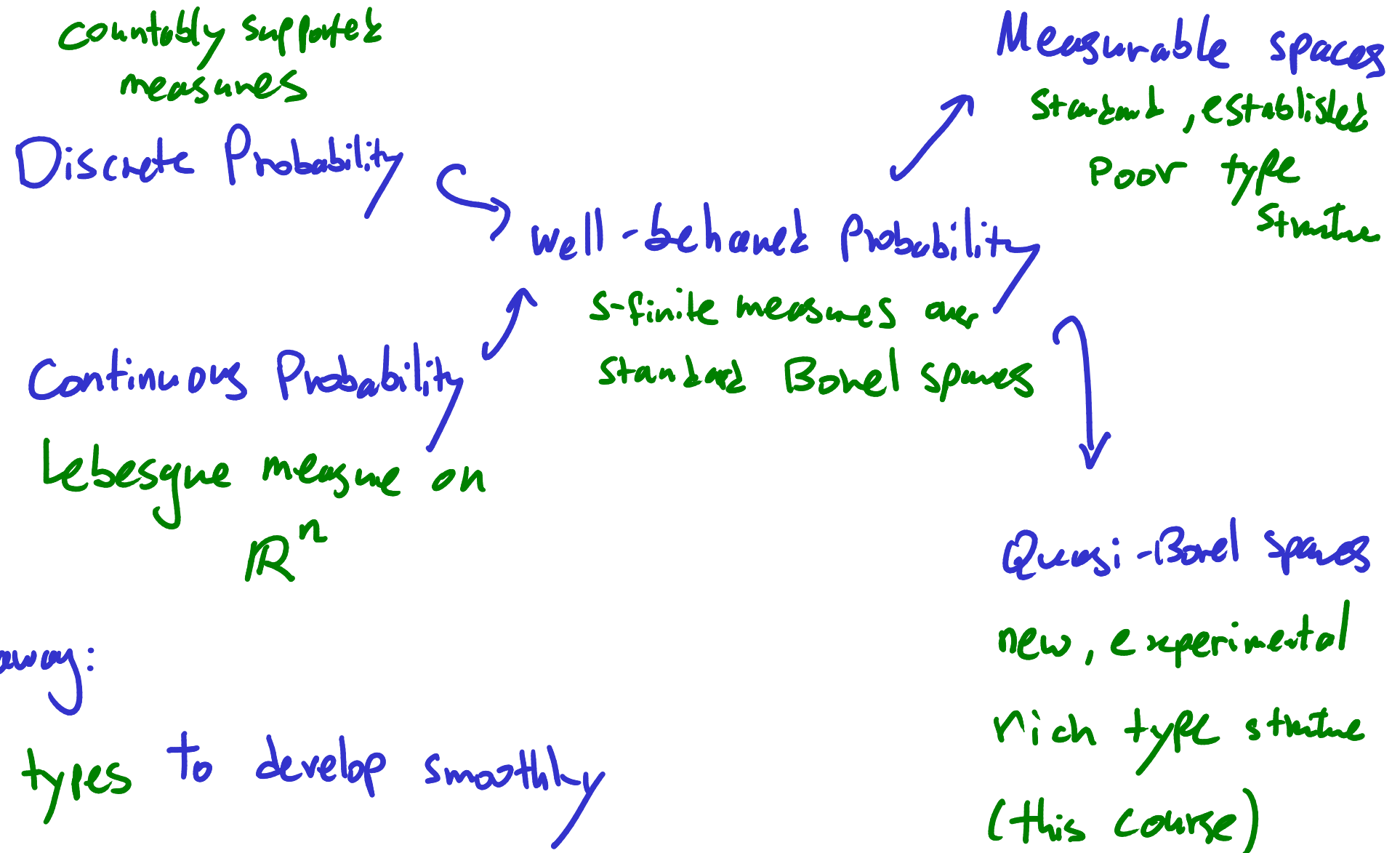
Clear connection to

Foundations:

- Reid's
- John's courses
- Michael's
- Dominik's

- this course

# Why foundations?



Takeaway:

use types to develop smoothly

Plan:

- 1) Type-driven probability: discrete case (Mon + Tue (?))
- 2) Borel sets & measurable spaces (Tue)
- 3) Quasi Borel spaces, simple type structure (Wed)
- 4) Dependent type structure & standard Borel spaces (Thu)
- 5) Integration & random variables (Fri)

please ask questions!



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# Language of distribution & Probability

$X$  type (=space) of values / outcomes

$DX$  type of distributions / measures over  $X$

$PX \subseteq DX$  Sub type of probability measures (total measure 1)

$BX$  type of measurable events - subsets of  $X$  we wish to measure

$W$  type of weights :  $[0, \infty]$

$\rightarrow$  type judgment

$\mu : DX, E : BX \vdash c_{\mu}[E] : W$

$\hookrightarrow$  measure  $\mu$  assigns to  $E$



# Axioms for measures

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Empty event:  $\emptyset : \mathcal{B}X$

Its measure is  $0 : \mathcal{W}$ :

$$\mu : \mathcal{D}X \vdash \underbrace{C_e[\emptyset]}_{\mu} = 0 : \mathcal{W}$$

# Axioms for measures

---

$\mathcal{B}X$  is a Boolean sub-algebra:

$$E : \mathcal{B}X \vdash E^c : \mathcal{B}X$$

$$E, F : \mathcal{B}X \vdash E \cup F, E \cap F : \mathcal{B}X$$

$E, C : \mathcal{B}X, \mu : \mathcal{D}X \vdash$  (disjoint additivity)

$$\mu[E] = \mu[E \cap C] + \mu[E \cap C^c] : \mathbb{W}$$

# Axioms for measures

$\omega := (\mathbb{N}, \leq)$      $(B, \subseteq)$      $(W, \leq)$  posets

$$(BX, \subseteq)^\omega := \left\{ (E_n)_{n \in \mathbb{N}} \in (BX)^\mathbb{N} \mid E_0 \subseteq E_1 \subseteq E_2 \subseteq \dots \right\}$$

$(BX, \subseteq)$  and  $(W, \leq)$  are  $\omega$ -chain-closed:

$$E_- : (BX, \subseteq)^\omega \vdash \bigcup_n E_n : BX \qquad a_- : (W, \leq)^\omega \vdash \sup_n a_n : W$$

$$E_- : (BX, \subseteq)^\omega, \mu : DX \vdash \qquad \text{(Scott continuity)}$$

$$C_e[\bigcup_n E_n] = \sup_n C_e[E_n] : W$$

# Axiom for Probability

$$\text{Cost} : PX \xrightarrow{\varepsilon} DX$$

$$1 : W$$

$$\mu : PX \vdash \underset{\text{Cost } \mu}{Ce[X]} = 1 : W$$

Avoid casting:

$$E : BX, \mu : PX \vdash \underset{\mu}{Pr[E]} := \underset{\text{Cost } \mu}{Ce[E]} : [0,1] \subseteq W$$

# Axioms for measures

Integration:

$$\mu: \mathcal{D}X, \varphi: \mathcal{W}^X \mapsto \int \mu \varphi : \mathcal{W} \quad (\text{Lebesgue integral})$$

Again, avoid casting:

$$\mu: \mathcal{P}X, \varphi: \mathcal{W}^X \mapsto \underbrace{E[\varphi]}_{\mu} := \int (\text{cast } \mu) \varphi : \mathcal{W} \quad (\text{Expectation})$$

More structure & notation later (...technical...)

Have: language + axioms

Want: model

today: discrete measures

rest of course: discrete + continuous

# Discrete model

type  $X$ : set

$$DX := \left\{ \mu: X \rightarrow \mathbb{W} \mid \mu \text{ is countably supported} \right\}$$

(next slide)

# Support

→ Powerset

$\mu: W^X$ ,  $S: \mathcal{P}X \vdash S$  supports  $\mu :=$

$\forall x: X. \mu x > 0 \Rightarrow x \in S$  : Prop

$\mu: W^X \vdash \text{supp } \mu := \{x \in X \mid \mu x > 0\} : \mathcal{P}X$

supp  $\mu$  is the smallest set supporting  $\mu$



# Discrete model

type  $X$ : set

$$DX := \{ \mu: X \rightarrow \mathbb{W} \mid \mu \text{ is countably supported} \}$$

$$:= \{ \mu: X \rightarrow \mathbb{W} \mid \text{supp } \mu \text{ is countable} \}$$

## Ex. measures

- $X$  ctbl, Counting measure  $\#_X : DX$

$$\#_X := \lambda x : X. 1 \quad (\text{NB: } \text{Supp} \#_X = X \quad \checkmark \text{ ctbl})$$

- Dirac measure:

$$x : X \mapsto \delta_x := \lambda x'. \begin{cases} x = x' : 1 \\ \text{o.w.} : 0 \end{cases} : DX$$

$$\text{NB: } \text{Supp} \delta_x = \{x\} \quad \checkmark \text{ ctbl}$$

- Zero measure  $\underline{0} := \lambda x. 0 : DX$

$$\text{NB: } \text{Supp} \underline{0} = \emptyset \quad \checkmark \text{ ctbl}$$

# Discrete model

type  $X$ : set

$DX := \{ \mu: X \rightarrow \mathbb{W} \mid \mu \text{ is countably supported} \}$

$$\mu: DX, E: \mathcal{B}X \vdash \underbrace{C_e[E]}_{\mu} := \sum_{x \in E} \mu x$$

$$:= \sum_{x \in E \cap \text{Supp } \mu} \mu x$$

Lemma:  $\mu: DX, S \in \mathcal{P}_{\text{ctbl}} X, S \text{ supports } \mu, E: \mathcal{B}X \vdash$

$$\underbrace{C_e[E]}_{\mu} = \sum_{x \in E \cap S} \mu x$$

Ex:

•  $E: \mathcal{B}X \vdash \quad C_e[E] = |E| := \begin{cases} E \text{ has } n \text{ elements: } n \\ E \text{ infinite: } \infty \end{cases}$   
 $\#_x$

•  $E: \mathcal{B}X, n: X \vdash \quad C_e[E] = \begin{cases} x \in E: 1 \\ x \notin E: 0 \end{cases} =: [x \in E] : \mathbb{W}$   
 $\delta_n$

NB:  $E: \mathcal{B}X \vdash [- \in E] : X \rightarrow \mathbb{W}$

indicator  
function

•  $E: \mathcal{B}X \vdash \quad C_e[E] = 0$   
0

# Validate axioms

$$\mu:DX \vdash C_{\mu}[E] = 0 \quad : W$$

$$E, C : BX, \mu:DX \vdash$$

$$C_{\mu}[E] = C_{\mu}[E \cap C] + C_{\mu}[E \cap C^c] \quad : W$$

$$E_{-} : (BX, \subseteq)^W, \mu:DX \vdash$$

$$C_{\mu}[\bigcup_n E_n] = \sup_n C_{\mu}[E_n] \quad : W$$

# Kernels

$\kappa$  from  $\Gamma$  to  $X$ :

$$\kappa : (\mathcal{D}X)^\Gamma$$

kernels are "open/parameterised" measures

Ex: Dirac kernel.  $\delta_x : (\mathcal{D}X)^X$

# Koçk Integral

$$\mu : D\Gamma, \kappa : DX^\Gamma \vdash \oint \mu \kappa : DX$$

In discrete model:

$$\oint \mu \kappa := \lambda x : X. \sum_{\nu \in \Gamma} \mu \nu \cdot \overbrace{k(\nu; x)}{:= k \nu x}$$

# (Weak) disintegration problem:

Input:  $\mu: D\Gamma$   $V: DX$

Output: a kernel  $k: (DX)^\Gamma$  s.t.

$$\int \mu k = V$$

Call such  $k$  a (weak) disintegration of  $V$

w.r.t.  $\mu$ .

(non-standard terminology)



Ex disintegration:

$$\underline{n} := \{0, 1, 2, \dots, n-1\}$$

disintegrate  $\#_{\underline{z}^{n+1}}$  w.r.t.  $\#_{\underline{z}}$

$$k: (D(\underline{z}^{n+1}))^{\underline{z}} \quad k(x; f) := \begin{cases} f(n) = x: & 1 \\ \text{o.w.} & : 0 \end{cases}$$

$$\left( \int \#_{\underline{z}} k \right) f = \sum_{x \in \underline{z}} \overbrace{\#_{\underline{z}}^1 x}^1 \cdot k(x; f)$$

NB:  $\text{Supp}(kx)$   
 $\sqrt{c + b}$

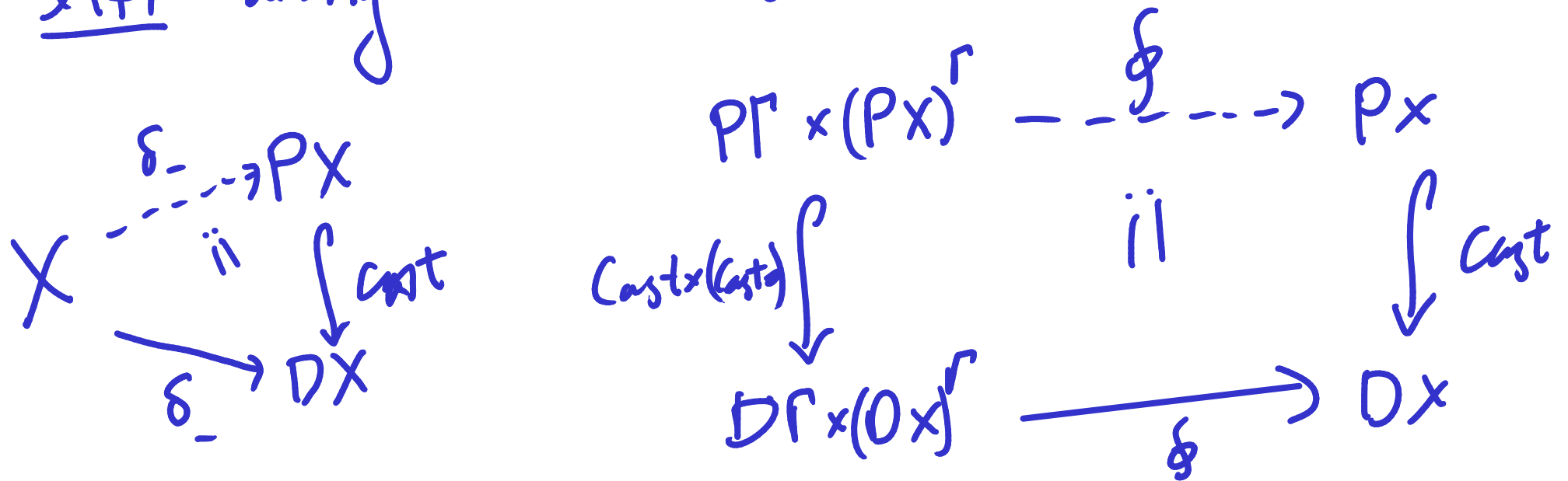
$$= k(0; f) + k(1; f) = k(fn; f) = 1 = \#_{\underline{z}^{n+1}}(f)$$

# Probability measures

$$PX := \{ \mu : DX \mid C_{\mu}[X] = 1 \} \xrightarrow{\subseteq} DX$$

Lemma:  $\delta_- : X \rightarrow DX$  and  $\wp : D\Gamma \times (DX)^{\Gamma} \rightarrow DX$

lift along the inclusion  $\text{cast} : P \xrightarrow{\subseteq} D :$



Prop (discrete Giry):

(Michèle Giry '82)

$(D, \delta_-, \oint)$  is a monad i.e.

$$\mu: \Gamma, \kappa: (DX)^\Gamma \vdash \oint \delta_\mu \kappa = \kappa \mu$$

$$\mu: DX \vdash \oint \mu(\delta x) \delta_x = \mu$$

: DX

$$\mu: D\Gamma, \kappa: (DX)^\Gamma, \tau: (DY)^X \vdash$$

$$\oint \mu(\delta \mu) (\oint (\kappa \tau) \tau) = \oint (\oint \mu \kappa) (\delta \tau) \tau(\alpha)$$

Corollary:  $(P, \delta_-, \oint)$  is a monad.

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