#### Foundations for type-driven probabilistic modelling

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Logic Summer School Australian National University 4-16 December, 2023 Canberra, ACT, Australia









THE ROYAL The
SOCIETY Alan Turing Facebook Research NCSC

Recop Language et distribution & Probability type (=space) of values/outcomes type of distributions/measures over X PX Sustype of Probability measures (total measure) type of measurable events - subsets of X we wish to measure type of weights: [0,00] W judgent M: Dx, E: BX+ Ce[E]: W Lo measure  $\mu$  assigns to E

## Arions for measures/distributions

Recop

$$\mu:DX + Ce[\emptyset] = 0 : W$$

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

$$Ce[E] = Ce[Enc] + Ce[Enc] : W$$

$$F = (BX, S), \mu:DX +$$

$$Ce[VEn] = Sup Ce[En] : W$$

Kernels & their koch integral

Kernel from I to X: k: (OX) or k: I -> DX Dirac kernel: S\_:X→DX Kock integnl: M:DT, k:(DX) + &Mk or f \( \( \dn \) \( \kappa \) \( \dn \) \( \d Giry monals: (D, S\_, \$) A (P, S\_, \$).

#### Discrete model

Type: Set 
$$W := [0,\infty]$$
  $Bx := Px$ 
 $DX := \{\mu : X \to W \mid Supp \mu \text{ Countable } \}$ 
 $PX := \{\mu \in OX \mid Ce[X] = 1\}$ 
 $Ce[E] := \sum_{n \in E} \mu_n S_n := \lambda n'. \begin{cases} n = n' : 0 \\ n \neq n' : 1 \end{cases}$ 
 $Px := \lambda n \cdot \sum_{n \in E} \mu_n \cdot k(r_i \times n_i)$ 

#### Ex distributions

Counting measure (xets): 
$$\# := \lambda x. 1$$

Zero measure 
$$o := \lambda x.0$$

### Plan:

- 1) type-driven probability: discrete case (Mon + Tue)
- 2) Borel sets & measurable spaces (Tue)
- 3) Quisi Borel spaces, simple type struitue (Web)
- 4) Dependent type structure & standard Barel Spines (Thus)
- 5) Integration & random variables (Fri)

Pleuse as n guestions!



y Course web page

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## Product measures

= 
$$\lambda(2,7)$$
.  $\mu z \cdot \nu y$   
1 discrete  
model

# (build measures) (compositionally)

#### Indeel:

Notation:  $\lambda: D(xxy), k(DZ) + \iint \lambda(dz, dy) k(z, y)$ := fxk

Fulini-Tonelli Thm:

Integrate in any order:

M:DX, V:DY, K:(DZ) 1-

 $\oint \mu(\lambda n) \oint V(\lambda y) \, u(\lambda, y) = \iint (\mu(\mathcal{D}) V(\lambda \lambda, \lambda y))$  $= \phi V(2y) \phi \mu(2x) k(n,y)$  Pushing a measure forward  $\mu:D_{\Omega}, \alpha: x^{\Omega} + \mu:= \phi \mu(\omega) S : DX$ = Xx. I YW WEST

a:X: random element

M:DX: the law of a

(wr.t. m)

En: We can represent configurations of 2 dice using 6×6 Letting (+): 62 -> 1N2 (+) we he that the law

(#60#6)(+) DIN

is the number of wolls whose sur is gim

$$(\cdot): \mathbb{W} \times \mathbb{D} \times \longrightarrow \mathbb{D} \times$$

(.): 
$$W \times DX \longrightarrow DX$$
  
 $\alpha \cdot \mu := \lambda \pi. \quad \alpha \cdot \mu \times$ 

$$NB: S-pp(\alpha \cdot \mu) = \begin{cases} \alpha = 0: & \beta \\ 0 \neq 0: & S-pp \mu \end{cases}$$

#### Normalisation

$$\mu:0x$$
,  $Ce[X] \neq 0,\infty +$ 

$$||\mu|| := (ce[X]) \cdot \mu : PX$$

$$E_X:$$

$$\phi \neq A \subseteq_{fin}^X : \overline{U} := \|(\#_A)\|_{A \subseteq X} : PX$$

$$1 \xrightarrow{\#_{A}} DA \xrightarrow{(-)_{A \leq X}} DX \xrightarrow{\|-\|} PX$$

I.e. 
$$U := \lambda n. \begin{cases} n \in A: \frac{1}{|A|} \\ n \notin A: 0 \end{cases}$$
 so  $U = S_n$   $\begin{cases} n \notin A: 0 \end{cases}$ 

## Stanbart Vocabulan

$$\mu: D(x \times x_2)$$

marginalisation: 
$$\mu_{\pi, =}$$
  $ff$   $\mu(d\pi, dy) S_{\infty}$ 

integrate out y

Exercise: 
$$\mu: PX, V: DX \vdash (\mu \otimes V)_{\pi_2} = V$$

 $X_1 \leftarrow X_1 \times X_2 \xrightarrow{X_2} X_2$ 

#### independence

Pairing r.e.s:

d: x 2, B: Y2 L

LX,B> = LW. LXW, FWS: (X+Y)

 $\lambda:D\Omega$ ,  $\alpha:x^{\Omega}$ ,  $\beta:Y^{\Omega}$  +  $\alpha\perp\beta:=\lambda$   $\alpha:=\lambda$   $\alpha:=\lambda$ 

or, & independent W.r.t. >

Ex represent outcomes of 3 coin tosses: C:={T,H} \( \alpha := C \times C \times C \times C \times U Ti: 12 -> C outcome of ith toss

Same:  $\Omega \xrightarrow{\langle \pi_i, \pi_i \rangle} C \times C \xrightarrow{(=)} B$ 

whee:  $(\stackrel{?}{=}):C^2 \rightarrow B:=\{True, False\}$   $n=y:=\{n=y:True, False\}$   $n\neq y:=\{n\neq y:False\}$ 

Ex represent outcomes of 3 coin tosses: C:={T,H} 2:= C \* C \* C 1: U&U&U : P1 Morginalisation  $V_{c}(T).v_{c}(T)$   $\lambda T = (V_{c} \otimes V_{c})T = \frac{1}{4} + \frac{1}{4} = \frac{1}{2}$   $\lambda Same_{12}$   $V_{c}(T).v_{c}(T)$   $V_{c}(T).v_{c}(T)$ 

Ex represent outcomes of 3 coin tosses: C:={T,H} 2:= C \* C \* C 1:U&U&U:P1 Ti: 12 -> C Outcome of ith toss Same: 12 (=) B i+j: \some :: B  $\lambda$ :  $(T,T)H = \frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{4}$   $(Same_{12}, Same_{23})$   $(T,T)H = \frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{4}$   $(Same_{12}, Same_{23})$  (T,T,T)(T,F) + 1.12-1 4) \(\(\mu, \mu, \tau, \tau\)

Ex represuit outcomes of 3 coin tosses: C:={T,H} 2:= C \* C \* C 1: U & U & U & C : P 1 Ti: S2 -> C Outcome of ith toss Same: 12 (=) B i+j: \some; B  $\lambda = U = U \otimes U = \lambda \otimes \lambda$   $\langle Same_{12}, Same_{12} \rangle = ||B| \times ||B|| = ||B| \times ||B||$ So Sane I Same
12 à 13

#### independence

Ex represuit outcomes of 3 coin tosses: C:={T,H} 2:= C \* C \* C 1: U&U,OU: P2 Ti: 12 -> C Outcome of ith toss Same:  $\Omega \xrightarrow{\langle \pi_i, \pi_i \rangle} C \times C \xrightarrow{(=)} B$ i # j: Some: j i + j: Same I Same jn 1 { Sane, Sa Intuition: Same 3 = IFF (Same, 2, Same, 23)
calc:  $\lambda \left( T, T, T \right) = \frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{4} \neq \frac{1}{2^3} = \lambda \otimes \lambda \otimes \lambda \otimes \lambda$ \( \lambda \text{Same}\_{12}, \text{Same}\_{12}, \text{Same}\_{13} \)
\( \lambda \text{A(H, H, H)} \)
\( \lambda \text{A(T, T, T)} \)
\[
\lambda \text{A(T, T, T)} \]

## Vocabulan

(Discrete) Measure space 
$$(X, \mu: DX)$$

Measure preserving  $f:(X,\mu) \longrightarrow (Y,\nu)$ 

furtion  $f: X \to Y$  s.t.  $M_f = V$ 
 $\mu: DX, f: X \to Y + \mu \text{ invariant under } f:=$ 
 $f:(X,\mu) \to (X,\mu)$ 

Ea:

 $\mu: DX, \nu: DY + \nu = \nu = \nu = \nu$ 

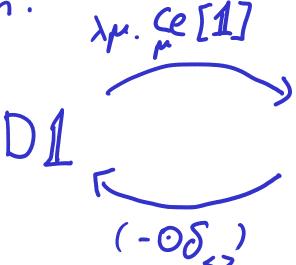
Swap: $(X \times Y, \mu \otimes V) \longrightarrow (Y \times X, \nu \otimes \mu)$  so

 $\mu: DX + \nu \otimes \mu = \nu = \nu = \nu = \nu = \nu$ 
 $\nu: DX + \nu \otimes \mu = \nu = \nu = \nu = \nu = \nu = \nu$ 

## Weights as mecsures

NO: unit relie

Observation:



mutually inverse

Pruf:  $\mu \mapsto \mu(<>)$ 

 $A\omega.a \longleftrightarrow A \alpha$   $(A\omega.a)c$ 



(Lebesghe ) integral

Can derive it:

$$DX \times W \xrightarrow{D \times \times (\stackrel{\sim}{=} \circ -)} Dx \times (D1)^{X}$$

$$:= \qquad \qquad \downarrow \phi$$

$$W \leftarrow \qquad \qquad D1$$

I ctb1, 
$$\mu:(DX)^T + \sum_{i \in I} \mu_i : DX$$

$$:= \lambda_{x}. \sum_{i \in I} \mu_{x}$$

Thm (affine-linearity):

of is affine-linear in each argumt:

J ctsl  $M: (O\Gamma)^T, k: (DX)^T + \oint (\sum a_i P_i) k = \sum a_i \oint P_i k$   $a_i \in \Gamma$   $i \in \Gamma$   $i \in \Gamma$ 

J (+51, M:DT, a:W, k:Dx +

$$\int \mu(dn) \left( \sum_{i \in I} a_i \cdot k_i(x) \right) = \sum_{i \in I} a_i \cdot \oint \mu k_i$$

Prop:  $W \cong D1$  is a 6-Seri-ring isomorphism:  $(W, \Sigma, (\cdot), 1) \cong (D1, \Sigma, (\cdot), S_s)$ and  $(\cdot): W \times DX \longrightarrow DX$  mong DX into a module:

$$(\Sigma_{\alpha_i}) \cdot \mu = \Sigma_{(\alpha_i, \mu)}$$
  $\alpha \cdot \Sigma_{i \in I} = \Sigma_{i \in I} \alpha \cdot \mu_i$ 

Cordlan: sis affine-linear in each argument.

## Randon Variable:

NB: 1R:=[-00,00]

A random element  $\alpha: \overline{R}^{\Omega}$  (wrt some  $\mu:D\Omega$ )

can add, multiply r.v.'s.

To integrate t.v.s:

$$(-)^{t}: \mathbb{R}^{n} \longrightarrow \mathbb{W}^{n}$$

$$\alpha := \lambda \omega . \begin{cases} \alpha \cdot \omega \geqslant 0 : \alpha \omega \\ 0 \cdot \omega : 0 \end{cases} = \left[ \alpha - \geqslant 0 \right] \cdot |\alpha|$$

$$\alpha = \lambda \omega \cdot \begin{cases} \alpha \cdot \omega \leq 0 : |\alpha \omega| \\ 0 \cdot \omega \cdot \vdots \\ 0 \end{cases} = [\alpha - \leq 0] \cdot |\alpha|$$

So 
$$\alpha = \alpha^{\dagger} - \alpha^{-}$$

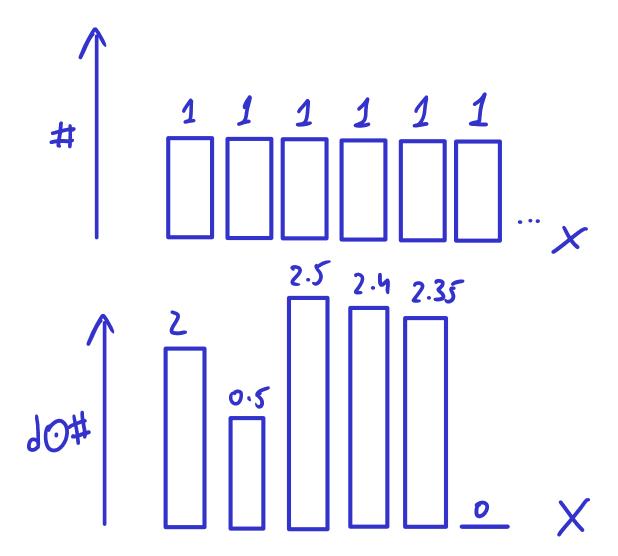
Density

a density over X: d:X->W

 $d:W^{\times}$ ,  $\mu:DX \vdash dO\mu$ : DX  $:= \phi \mu(un)(dn \cdot S_n)$ 

Warning The types of measures & densities in the discrete model are close, but still different. They wincide on countable sets, so people often confue them. Types help us keep them separate.

#### Intuition:



## Almost certain Properties

E:BX, 
$$\mu$$
:DX +  $\mu(m)$ -almost certainly  $x \in E$ : Prop

:=  $[-EE]O\mu = \mu$ 
 $1 \times [nEE] = \begin{cases} n \in E : I \\ n \notin E : O \end{cases}$ :W

When p: Px we say instend

p(dn)-almost Surely x EE

Exercise Look up the def. of a normel space and mobility the definition so that  $I_p(\Omega, \mu)$  is a normel space up to almost sure equality.

Absolute continuity

d is a density of p w.r.t. V or d is a Radon-Nikodym derivative wit. V

 $\mu, \nu: Dx, d: W^{X} + d = \frac{d\mu}{d\nu}$ 

: Prop

:= \mu = dov

=: pu has a density w.r.t. V

Lemma: perv,
h: (py)x

 $\int V(dx) \frac{d\mu}{dv}(x) \cdot kx = \int \mu(dx) kx$ 

but also: 
$$\frac{dU_{ASX}}{d(\ddagger_{A'cont})} = \lambda x \cdot \frac{1}{1A1}$$

Radon-Nikodym Thm: (discrete version)

p,v:Px+ p«v iff Vn. Vx=0 => px=0

ie. Supp p = Suppv

In that case, if  $d_1, d_2 = \frac{d\mu}{d\nu}$  then Y(dn) - a.s.  $d_1 z = d_2 z$ 

En: for dbl X,  $\forall \mu:0x$ .  $\mu\ll\#_X$ . Proof: Vacuously, as  $\#_X n\neq 0$ .

Thun  $\exists x. \mu x = \frac{d\mu}{d\#}$ .

#### Consitional expectation

B is a conditional expectation of a wrt. µ along H μ:002, H: x<sup>2</sup>, d: [(Ω,μ), β: [(X, r<sub>H</sub>) + B = 1 [ \ H = - ]  $:= \forall \varphi: \int_{\Gamma_{i}} (Y_{i}, Y_{H}) \cdot \left( \prod_{i=1}^{n} (Y_{i}, Y_{H}) \cdot (Y_{H}) \cdot ($ Perfect X Statistic

Thm (Holmogotov): (discrete version)

There is a funtion

$$\mathbb{E}\left[-|-=-\right] \in \prod \prod \prod \int_{\mu: P_{\Omega}} \mathcal{L}_{1}(\Omega, \mu) - \mathcal{L}_{1}(X, \mu_{H})$$

S.t. IE[a|H=-] is a conditional eapertation of a wirt pre along H.

Bayes's The (discrete version, adopted for Williams): Let  $\lambda: P(X \times \Theta)$  joint probability distribution.

Assume M: OX V:DED St. DEC MEDV.

Assure  $\mu: D \times , V:D \oplus S.t. \lambda \ll \mu \otimes V.$ with  $d_{X,H} = \frac{d\lambda}{d(\mu \otimes V)}$ .

Obs 1: 
$$d_x:W^X$$
 then  $d_x = \frac{d\lambda_0}{d\mu}$   
 $d_x:=\lambda_x. \int V(\lambda_0) d(\lambda_1,0)$ 

$$A \leq \min_{x \in A} d(\lambda_1) = \lambda_0 \int (\lambda_1(\lambda_1) + \lambda_1(\lambda_2)) d(\lambda_1) = \frac{d\lambda_0}{d\mu}$$

A similary 
$$(d_{14}:W^{\Theta}):=\lambda\Theta.\int \mu(dx)d(2,8)=\frac{d\lambda_{H}}{d\nu}$$

XXA)

H:= n<sub>2</sub>

X

A

Obsaulie

Bayes's Thm (discrete version, adopted for Williams):

Let 
$$\lambda: P(X \times M)$$
 joint probability distribution.

Assure 
$$\mu: DX, V: D\Theta$$
 S.t.  $\lambda \ll \mu \otimes V$ .

with 
$$d_{X,H} = \frac{d\lambda}{d(\mu \otimes v)}$$
 .  $d_{X} = \frac{d\lambda_{H}}{d\mu}$   $d_{X} = \frac{d\lambda_{H}}{d\nu}$ 

$$= \frac{d\lambda}{d(\mu \otimes \nu)} \cdot dx = \frac{d\lambda}{d\mu}$$

$$d = \frac{d\lambda_H}{d\nu}$$

Let 
$$d_{X/H}(-1-):X\times O \longrightarrow W$$

$$\lambda_{x|H=-}$$
:  $\Theta \rightarrow PX$ 

Summay Product measures 4 Fubini-Tonelli MOV Push - forward / law MH module structure ont affine linearity of 9  $(Dx, \mathcal{I}, (\cdot))$ Lebesque integration Student vocabulog: joint dist., morginalisation, independence, invariance density & Radon - Nikodym derivatives (heed the Worning) almost certain properties Conditional expectation A Probability with Bayes's Thm.

Plan:

1) type-driven probability: discrete case (Mon + Tue)
2) Borel sets & measurable spaces (Tide)

3) Quisi Borel Spaces, Single type struitue (Web)

4) Dependent type structure & standard Barel Spines (Thus)

5) Integration & random variables (Fri)

Pleuse as n guestions!

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Course web page