

Thm 4.3 converse to univariate Mecke:

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Let  $\eta$  be a  $\sigma$ -finite measure on  $(X, \mathcal{X})$  and  $\eta$  is a pt. process on  $X$  satisfying

$$E \int_X f(x, \eta) \eta(dx) = \int_X E f(x, \eta + \delta_x) \lambda(dx) \quad *$$

$\forall f \in \mathbb{R}_+(X \times \mathcal{N})$ , then  $\eta$  is

a PPP with mean measure  $\lambda$ .

Ex Given  $k > 1$ , prove or disprove that  $\eta$  is a converse to  $k$ -variate Mecke

Proof Let  $A_1, \dots, A_m$  be disjoint sets in  $\mathcal{X}$

with  $\lambda(A_i) < \infty$ . Let  $k_1, \dots, k_m \in \mathbb{N}_0$ . Let

$$f(x, \eta) = \mathbb{1}_{\{x \in A_1, \eta(A_1) = k_1, \dots, \eta(A_m) = k_m\}}$$

$$\text{Then } E \int_X f(x, \eta) \eta(dx) = E \left[ \eta(A_1) \mathbb{1}_{\{\eta(A_1) = k_1, \dots, \eta(A_m) = k_m\}} \right]$$

$$= k_1 P[\eta(A_1) = k_1, \dots, \eta(A_m) = k_m]$$

$$E f(x, \eta + \delta_x) = \mathbb{1}_{A_1}(x) P[\eta(A_1) = k_1 - 1, \eta(A_2) = k_2, \dots, \eta(A_m) = k_m]$$

so by (\*),

$$\begin{aligned} & k_1 P[\eta(A_1) = k_1, \dots, \eta(A_m) = k_m] \\ &= \lambda(A_1) P[\eta(A_1) \leq k_1 - 1, \eta(A_2) = k_2, \dots, \eta(A_m) = k_m] \end{aligned}$$

so putting  $\pi(k) = P[\eta(A_1) = k \mid \bigcap_{i=2}^m \{\eta(A_i) = k_i\}]$

$$\pi(k) = \lambda(A_1) \pi(k-1) / k \quad \text{so}$$

$$\pi(n) = \pi(0) \frac{\pi(1)}{\pi(0)} \cdots \frac{\pi(n)}{\pi(n-1)} = \pi(0) \frac{\lambda(A_1)^n}{n!}$$

so  $P[\eta(A_1) \mid \bigcap_{i=2}^m \eta(A_i) = k_i] \sim P_0(\lambda(A_1))$

regardless of  $k_2, \dots, k_m$  so  $\eta(A_1) \sim P_0(\lambda(A_1))$ ,

$$\eta(A_1) \cdots \eta(A_m)$$

By induction on  $m$ ,

$$\eta(A_1) \cdots \eta(A_m) \text{ indep.}$$

If  $\lambda(A) = \infty$  can write  $A = \bigcup_{i=1}^{\infty} A_i$

with  $A_i$  disjoint,  $\lambda(A_i) < \infty$ ,  $\sum_{i=1}^{\infty} \lambda(A_i) = \lambda(A) = \infty$ .

$$\begin{aligned}
P[\eta(A) \leq k] &\leq P[\eta(\bigcup_{i=1}^n A_i) \leq k] \\
&= \sum_{j=1}^k P_0(\sum_{i=1}^n \lambda(A_i) \leq j) \\
&\rightarrow 0 \text{ as } n \rightarrow \infty
\end{aligned}$$

so  $\eta(A) = \infty$  a.s.

Prop 4.4 Suppose  $N$  is a  $N_0$ -valued RV and  $\gamma > 0$ . Then  $\nu \sim P_0(\gamma)$  iff  $\forall f \in B_b(N_0)$

$$E[N f(N)] = \gamma f(N+1), \quad (*)$$

Proof Direct:  $\Rightarrow$  ex  $\Leftarrow$  take  $f(k) = \mathbb{1}_{\{x=k\}}$ .

Special case of Thms 4.1 & 4.3. Take  $X = \{$ -point set  $x = \{x\}$ . Then  $\eta(x)$  is just a RV  $N \sim f(x, \eta) = g(N)$

$$\int f(x, \eta) \eta(dx) = N g(N)$$

$$\int E f(x, \eta + \delta_x) \lambda(dx) = \gamma E g(N+1)$$

Stein char: if  $(*)$  approx holds,  $N$  must be approx.  $P_0(\gamma)$ .

## 5 Mappings, Markings & Thinnings

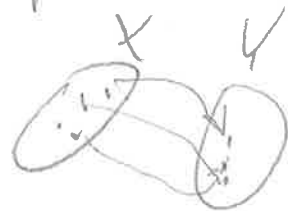
(75)

(operations that preserve PPs)

Suppose  $(X, \mathcal{X})$  &  $(Y, \mathcal{Y})$  are measurable spaces

&  $T: X \rightarrow Y$  is measurable. Given pt process

$$\eta = \sum_{i=1}^K \delta_{x_i} \quad \text{in } X$$



Let  $\tau(\eta) = \sum_{i=1}^K \delta_{T(x_i)}$  pt process in  $Y$

$$\tau(\eta)(A) = \eta(T^{-1}(A)) \quad A \in \mathcal{Y}$$

i.e.  $\tau(\eta) := \eta \circ T^{-1}$ . Use this defn for any pt pi.

Thm 5.1 Suppose  $\eta$  is a pt pr. on  $X$  with mean measure  $\lambda$ . Assume the measure

$T(\lambda) := \lambda \circ T^{-1}$  is  $\sigma$ -finite. Then

(i)  $\tau(\eta)$  is a pt pr in  $Y$  with mean measure  $T(\lambda)$

(ii) If  $\eta$  is a PPP, so is  $\tau(\eta)$ .

Proof (ii) For  $A \in \mathcal{Y}$ ,  $A_i \in \mathcal{J}$  disjoint. (76)

$$T(\eta)(A) = \eta(T^{-1}(A)) \in \mathbb{N}_0 \cup \{\infty\}$$

$$T(\eta)\left(\bigcup_{i=1}^{\infty} A_i\right) = \eta\left(T^{-1}\left(\bigcup_{i=1}^{\infty} A_i\right)\right)$$

$$= \dots = \sum_{i=1}^{\infty} \eta\left(T^{-1}(A_i)\right)$$

$$= \sum_{i=1}^{\infty} T(\eta)(A_i) \quad \text{disj.} \quad \text{so } T(\eta) \text{ is } \sigma\text{-finite}$$

$$\text{Also } E[T(\eta)(A)] = E[\eta(T^{-1}(A))] = \lambda(T^{-1}(A))$$

By Assumption  $\exists A_i$  disjoint with  $\bigcup_{i=1}^{\infty} A_i = Y$

$$\text{and } \lambda(T^{-1}(A_i)) < \infty$$

$$\text{so } \eta(T^{-1}(A_i)) < \infty \text{ a.s.}$$

$\eta \circ T^{-1}$  is  $\sigma$ -finite a.s.

Finally if  $B_1, \dots, B_n \in \mathcal{Y}$  are disjoint then

$T^{-1}(B_1), \dots, T^{-1}(B_n)$  are disjoint so

$\eta(T^{-1}(B_1)), \dots, \eta(T^{-1}(B_n))$  are indep Poissons

Now consider a marked pt process  
given  $\gamma = \sum_{i=1}^K \delta_{\xi_i}$  in  $X$

Suppose each pt.  $\xi_i$  has  
a  $Y$ -valued mark attached  $Y_i$



$P(Y_i | \gamma)$  depends only on  $\xi_i$ , i.e.  $P[Y_i \in dy | \gamma] = K(\xi_i, dy)$

Then  $\theta := \sum_{i=1}^K \delta_{(\xi_i, Y_i)}$  is a Marked pt p.f.

Formally suppose  $K$  is a stochastic kernel

i.e.  $K: X \times Y \rightarrow [0, 1]$  satisfies

- (i)  $K(x, \cdot)$  is a Prob. measure  $\forall x \in X$
- (ii)  $K(x, A)$  is  $\mathcal{X}$ -measurable on  $X, \forall A \in \mathcal{Y}$

Assume  $\gamma$  takes the form  $\gamma = \sum_{i=1}^K \delta_{\xi_i}$  and for

$k \in \mathbb{N}$  let  $P_k$  be a choice of the prob. dist of  $(\xi_1, \xi_2, \dots, \xi_k)$

Given  $K = k$  so  $P_k$  is a prob dist on  $X^k$

Define  $\tilde{P}_k$  on  $(X \times Y)^k$  by

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$$\tilde{P}_k(\uparrow) = \int \delta(x_1, \dots, x_n, y_1, \dots, y_n)$$

$$k(x_1, dy_1) \cdots k(x_n, dy_n) P_k(dx_1, \dots, dx_n)$$

(similar for  $k = \infty$ ). Put

$$\theta = \sum_{i=1}^K \delta_{(x_i, y_i)} \quad \text{where}$$

$$P_{\xi_1, \dots, \xi_K, \eta_1, \dots, \eta_K} (dx_1, \dots, dx_n, dy_1, \dots, dy_n | K=k) \\ = \tilde{P}_k (dx_1, \dots, dx_n, dy_1, \dots, dy_n)$$

Theorem 4.2 Given  $\theta$  as above,

$$L_\theta(u) = E e^{-\theta(u)} = L_\eta(u^*) \quad u \in \mathbb{R}_+(X \times Y)$$

$$\text{where } u^*(x) = -\log \int e^{-u(x,y)} k(x, dy)$$

Hence distribution of  $\theta$  does not depend on the

choice of  $\{\xi_i\}$ ,

$$\text{Proof } L_\theta(u) = \sum_{k \in [0, \infty]} P[K=k] \int e^{-\sum_{i=1}^k u(x_i, y_i)} k(x_1, dy_1) \cdots k(x_n, dy_n) P_k(dx_1, \dots, dx_n)$$

$$= \sum_{k \in [0, \infty)} P[K=k] \int \mathbb{P}_k (dx_1, \dots, dx_n)$$

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$$\times \prod_{i=1}^k e^{-u^*(x_i)}$$

$$= E \left[ e^{-\sum_{i=1}^K u^*(x_i)} \right] = L_{\eta}(u^*)$$

Thm 5.3 Suppose  $\eta$  is a PPP with  $\sigma$ -finite mean measure  $\lambda$ , then  $\theta$  (as above) is a PPP with mean measure  $\lambda \otimes \kappa$ :

$$\lambda \otimes \kappa (b) = \int \int b(x, y) \kappa(y) \lambda(dx)$$

$$b \in \mathbb{R}_+(X \times Y).$$

Proof Suppose  $\lambda(X) < \infty$ . Then may assume

$$\eta = \sum_{i=1}^K \delta_{x_i} \quad \kappa = \rho_0(Y), \quad \lambda = \lambda(X)$$

$$P[X_i \in C] = \frac{\lambda(C)}{\lambda} = \rho_0(C)$$

$$\theta = \sum_{i=1}^K \delta_{(x_i, Y_i)}$$

where  $(X_i, Y_i) \sim \mathbb{Q} \times K$  are indep,  $\mathbb{E} \neq 0$

hence by pt of Th 4.1

$$\Theta \text{ is a PPP intensity } \lambda \text{ on } \mathbb{Q} \times K \\ = \lambda \times K$$

suppose  $d(x) = \infty$  write  $d = \sum_{i=1}^{\infty} \lambda_i$   $d_i(x) = \infty$

and  $\eta = \sum_{i=1}^{\infty} \eta_i$ ,  $\eta_i$  PPP with  $d_i$  as mean meas

$\eta_i$  indep.

putting  $\theta_i$  the  $K$ -marked pp. associated

with  $\eta_i$

$$\Theta = \sum_{i=1}^{\infty} \theta_i$$

PPP with intensity  $\sum_{i=1}^{\infty} \lambda_i \otimes K$

$$= \lambda \otimes K.$$