

(1) 3/1/2013

Discrete Determinantal Point Process

Warm-up ($q=1$)

Y ... tall skinny orthogonal matrix

$$n \times p,$$

$$Y^T Y = I_p$$

$K = Y Y^T$ projection with eigs $1^p, 0^{n-p}$

hence positive semi-def

$$\text{Cauchy Binet } \det I_p = I_p(1 \dots p) = \sum_{i_1, \dots, i_p} Y(1 \dots p)_{i_1 \dots i_p}^2 = 1 \quad \left[\text{Tr } K = p \right]$$

$$K(i_1 \dots i_p) = \sum_{j_1, \dots, j_p} Y(i_1 \dots i_p)_{j_1 \dots j_p}^2$$

Two Discrete Probabilities

a) Prob on $\{1, \dots, n\}$

$$\text{Pr}(i) = \frac{1}{p} K_{ii} = \frac{1}{p} \sum_{j=1}^p Y_{ij}^2$$

b) Prop on p element subsets of $\{1, \dots, n\}$

$$\text{Pr}(i_1, \dots, i_p) = Y(i_1 \dots i_p)_{1 \dots p}^2 = K(i_1 \dots i_p)$$

= det of $p \times p$ principal submatrix of K
obtained by taking rows & cols in $\{i_1, \dots, i_p\}$

What is the connection?

Answer: Take a random number uniformly

from i_1, \dots, i_p generated from the second
and obtain the first.

(2) 3/1/2013

Proof,

$$\det(Y^T(I + K(z_1, \dots, z_n))Y) = 1 + \sum_{i \in \{1, \dots, n\}} z_i y \binom{y-i}{1-p}^2 + \dots$$

e.g. (coef of z_1) $\sum_{i \in \{1, \dots, p\}} y \binom{y-i}{1-p}^2 = \Pr(i \text{ shows up})$

$$\Pr(i) = \frac{1}{p} \sum_{i \in \{1, \dots, p\}} y \binom{y-i}{1-p}^2 = \Pr(i \text{ is chosen})$$

$$\det(I + K(z_1, \dots, z_n)) = 1 + \sum z_i K_{ii} + \dots \quad (1)$$

$$= \sum_{\substack{A \subseteq \{1, \dots, n\} \\ |A| \leq p}} z_A K \binom{A}{A} \leftarrow$$

can omit
but larger
are 0

Thus $\Pr(i) = \frac{1}{p} K_{ii}$ as defined in (a)

z_A takes column indices
from A in $K(z_1, \dots)$
and the rest from I

(3)

The bigger story

$$1 \leq q \leq p \quad (q \text{ fixed})$$

$\{c_{11}, \dots, c_{pp}\}$ generated with prob $K \binom{c_{ii} - c_{pp}}{c_{ii} - c_{pp}}$ as before
call it $\mathcal{J}_q \in \mathcal{S}_{1,1}^p$ of size $\binom{p}{2}$

Now pick q at random uniformly ($\text{Pr} = 1/\binom{p}{2}$)

Theorem: $\text{Pr}(\mathcal{E} | \mathcal{J}_q) = \frac{1}{\binom{p}{2}} K \binom{\mathcal{J}_q}{\mathcal{J}_q}$

Need to show $\sum_{\mathcal{J}_q \in \mathcal{S}_{1,1}^p} K \binom{c_{ii} - c_{pp}}{c_{ii} - c_{pp}} = K \binom{\mathcal{J}_q}{\mathcal{J}_q}$

since $\frac{1}{\binom{p}{2}} \sum_{\mathcal{J}_q \in \mathcal{S}_{1,1}^p} K \binom{c_{ii} - c_{pp}}{c_{ii} - c_{pp}} =$ the prob that \mathcal{J}_q is picked at random from $\mathcal{S}_{1,1}^p$

but obviously coef of expansion (1)

of \mathcal{J}_q

Focus on pairs $\text{Pr}(\bar{c}_{ij}) = \frac{1}{\binom{p}{2}} [K_{ii} K_{jj} - K_{ij}^2]$

$\text{Pr}(\bar{c}_{ij}) = \frac{1}{p(p-1)} [K_{ii} K_{jj} - K_{ij}^2] \approx \frac{1}{p^2} [K_{ii} K_{jj} - K_{ij}^2]$
where K_{ij} is like a $1 - \frac{K_{ij}^2}{K_{ii} K_{jj}}$ is like a correlation

④

$n \rightarrow$ continuous

$$Y = [\phi_0 \dots \phi_{p-1}] \quad \text{where } \int_{\text{interval}} \phi_i(x) \phi_j(x) dx = \delta_{ij}$$

$$Y^T Y = I_p$$

$$Y Y^T = K \quad K(x, y) = \sum_{i=0}^{p-1} \phi_i(x) \phi_i(y) \quad \leftarrow \text{rank } p$$

$$K(x, x) = \frac{1}{p} \sum_{i=0}^{p-1} \phi_i^2(x) \quad \text{is a prob density}$$

$$\int \int_{x_1, \dots, x_p} K \begin{pmatrix} x_1 & \dots & x_p \\ x_1 & \dots & x_p \end{pmatrix} dx_1 \dots dx_p = 1$$

or

$$\frac{1}{p!} \int \int_{\mathbb{R}^p} K \begin{pmatrix} x_1 & \dots & x_p \\ x_1 & \dots & x_p \end{pmatrix} dx_1 \dots dx_p = 1$$

Two are related

Pick $\{x_1, \dots, x_p\}$ randomly
then pick are uniformly
or a subset uniformly

5

Special case of orthogonal polynomial ensembles

Start with weight $w(x)$ ~~and interval~~ on \mathbb{R} (or subinterval)

$$\phi_i(x) = \prod_{k=0}^{i-1} (x - x_k) \sqrt{w(x)} = \left(\underbrace{c_i}_{\substack{\uparrow \\ \text{orthogonal} \\ \text{polynomial of degree } i}} (x) \underbrace{\sqrt{w(x)}}_{\substack{\uparrow \\ \text{so arbitrary} \\ \text{constant in} \\ \text{weight}}} \right)$$

Sometimes nice to make it a probability measure

$$K \begin{pmatrix} x_1 & \dots & x_p \\ x_1 & \dots & x_p \end{pmatrix} = \begin{vmatrix} \phi_0(x_1) & \dots & \phi_{p-1}(x_1) \\ \vdots & & \vdots \\ \phi_0(x_p) & \dots & \phi_{p-1}(x_p) \end{vmatrix}$$

$$= \prod_{k=0}^{p-1} (\text{leading coef of } \phi_k) \prod_{i < j}^2 |x_i - x_j| \prod_{i=1}^p w(x_i)$$

$$\begin{pmatrix} a_1 & b_1 & & & \\ & b_1 & & & \\ & & \ddots & & \\ & & & a_p & b_p \\ & & & & b_p \end{pmatrix} \begin{pmatrix} \phi_0 \\ \vdots \\ \phi_p \end{pmatrix} = \gamma \begin{pmatrix} \phi_0 \\ \vdots \\ \phi_p \end{pmatrix} = b_p \begin{pmatrix} 0 \\ \vdots \\ 0 \\ \phi_{p-1} \end{pmatrix}$$

$$\prod_{k=0}^{p-1} (\text{leading coef}) = b_1 b_2 \dots b_{p-1}$$

(6)

Also

$$\det(Y^T(I + K(z_1, \dots, z_r))Y) = \det(I + K(z_1, \dots, z_r))$$

$$= \sum_{i \in \{1, \dots, p\}} \prod (1 + z_i) \chi_{i, \dots, i}^2$$

$$= E\left(\prod (1 + z_i)\right)$$

e.g. Take $z_1 = \dots = z_p = \theta$ rest ~~0~~ -1
 get Prob all ~~0~~ $< r$

$$= \det(I - K|_{\theta, r})$$

Continuous $\det(I - \left[\int_{x \leq t} \phi_i(x) \phi_i(x) dx \right]) = \Pr[\text{all } d_i \leq t]$
 $= F_{\lambda_{\max}}(t)$

$$\det(I - K\chi_{x \leq t}) = F_{\lambda_{\max}}(t)$$

$K\chi_{x \leq t}$ is an operator that takes

$$f(y) \mapsto \int_0^{\infty} \left(\sum \phi_i(x) \phi_i(y) \right) |f(y)| dy$$