

The FKG inequality (I)

log-supermodular: (L, \leq) is a finite distributive lattice (f.d. lattice).

$\mu: L \rightarrow \mathbb{R}_{\geq 0}$ is log-supermodular if $\mu(x)\mu(y) \leq \mu(x \vee y)\mu(x \wedge y)$.

increasing (decreasing): $f: L \rightarrow \mathbb{R}_{\geq 0}$ is $\uparrow (\downarrow)$ if $x \leq y \Rightarrow f(x) \leq f(y)$ (\geq)

Thm If (L, \leq) is a f.d. lattice & $\mu: L \rightarrow \mathbb{R}_{\geq 0}$ is bg-supermodular,
then for any two \uparrow functions $f, g: L \rightarrow \mathbb{R}_{\geq 0}$ we have

$$\left(\sum_{x \in L} \mu(x)f(x) \right) \left(\sum_{x \in L} \mu(x)g(x) \right) \leq \left(\sum_{x \in L} \mu(x)f(x)g(x) \right) \left(\sum_{x \in L} \mu(x) \right)$$

Pf: Define $\alpha, \beta, \gamma, \delta: L \rightarrow \mathbb{R}_{\geq 0}$ s.t. $\alpha = \mu f$, $\beta = \mu g$, $\gamma = \mu fg$, $\delta = \mu$.

For $x, y \in L$, we have

$$\alpha(x)\beta(y) = \mu(x)f(x)\mu(y)g(y) \stackrel{(\because \text{log-supermodular})}{\leq} \underline{\mu(x \vee y)} \underline{\mu(x \wedge y)} \underline{f(x)g(y)} \stackrel{(\because f, g \uparrow)}{\leq} \gamma(x \vee y)\delta(x \wedge y).$$

HAT \Rightarrow for $\forall X, Y \subseteq L$, $\alpha(X)\beta(Y) \leq \gamma(X \vee Y)\delta(X \wedge Y)$

$$\Rightarrow \alpha(L)\beta(L) \leq \gamma(L)\delta(L) \quad \text{done!}$$

QED

The FKG inequality (II)

Corollary 1 If $f \downarrow$ & $g \downarrow$ on L , then FKG inequality also holds.

Pf: In the proof of FKG ineq., let $\alpha = \mu f$, $\beta = \mu g$, $r = \mu$ & $s = \mu fg$

To get $\alpha(x)\beta(x) \stackrel{\text{log-supersmodular}}{\leq} \mu(x \vee y) f(x) \mu(x \wedge y) g(x) \stackrel{f, g \downarrow}{\leq} \mu(x \vee y) s(x \wedge y)$.

QED

Corollary 2 If $f \uparrow$ & $g \downarrow$ (or vice versa), then in FKG inequality, the opposite inequality holds.

Pf: Let $k = \max_{x \in L} g(x)$. Then we have $f \uparrow$ & $k-g \uparrow$ from L to $\mathbb{R}_{\geq 0}$.

$$\text{FKG ineq.} \Rightarrow [\mu f(L)] [\mu(k-g)(L)] \leq [\mu f(k-g)(L)] [\mu(L)]$$

$$\Rightarrow [\mu f] [\mu k - \mu g] \leq [\mu f k - \mu f g] \mu$$

$$\Rightarrow \underbrace{[\mu f] [\mu k]}_{\text{omit } L} - \underbrace{[\mu f k] [\mu]}_{\mu} \leq [\mu f] [\mu g] - [\mu f g] \mu$$

$$\Rightarrow [(\mu f)(L)] [(\overset{\circ}{\mu g})(L)] \geq [(\mu f g)(L)] [\mu(L)].$$

QED

View μ as a measure on L

Thm If L is a finite distributive lattice and $\mu: L \rightarrow \mathbb{R}_{\geq 0}$ is a log-supermodular function which is not a zero function. Then for any $f \uparrow \& g \uparrow: L \rightarrow \mathbb{R}_{\geq 0}$, we have

$$\mathcal{E}(fg) \geq (\mathcal{E}f)(\mathcal{E}g)$$

where $\mathcal{E}f = \frac{\sum_{x \in L} f(x) \mu(x)}{\mu(L)}$.

Remark: This thm demonstrates the probabilistic nature of the FKG ineq.

The Harris-Kleitman inequality

- $\mathcal{A} \in 2^{\Omega}$ is a **downset (upset)** if $B \subset A \in \mathcal{A}$ ($B \supset A \in \mathcal{A}$) $\Rightarrow B \in \mathcal{A}$.
- $(\Omega, 2^\Omega, P)$ is a prob. space s.t. $P(\{A\}) = \prod_{i \in A} p_i \prod_{j \in \bar{A}} (1-p_j)$, where $0 \leq p_i \leq 1$
and $\Omega = 2^{[n]}$ (w.l.o.g.).

Thm For any $0 \leq p_i \leq 1$, $i=1, 2, \dots, n$, and $\mathcal{A}, \mathcal{B} \subseteq \Omega$.

$$\mathcal{A} \uparrow \text{ and } \mathcal{B} \uparrow \Rightarrow P(\mathcal{A} \cap \mathcal{B}) \stackrel{\text{usual intersection}}{\geq} P(\mathcal{A})P(\mathcal{B})$$

positively correlated
 negatively
 positively

\uparrow upset in 2^Ω \uparrow
 \downarrow downset in 2^Ω \downarrow

pf: (Ω, \leq) is a finite distributive lattice. $\mu(A) \stackrel{\text{def}}{=} P(\{A\}) \forall A \in \Omega$.

Let I_A, I_B be two indicate funcs. on Ω .

Claim: μ is log-supermodular. pf: By the fact that $(\prod_{i \in A} p_i)(\prod_{i \in B} p_i) = (\prod_{i \in A \cup B} p_i)(\prod_{i \in A \cap B} p_i)$.

Claim: I_A and I_B are \uparrow in the lattice (Ω, \leq) . pf: By the fact that $\mathcal{A} \uparrow$ and $\mathcal{B} \uparrow$.

$$P(\mathcal{A} \cap \mathcal{B}) = \left(\sum_{A \in \Omega} \mu(A) I_{\mathcal{A}}(A) I_{\mathcal{B}}(A) \right) \left(\sum_{A \in \Omega} \mu(A) \right)$$

$$\geq \left(\sum_{A \in \Omega} \mu(A) I_{\mathcal{A}}(A) \right) \left(\sum_{A \in \Omega} \mu(A) I_{\mathcal{B}}(A) \right) = P(\mathcal{A})P(\mathcal{B}).$$

QED

The Janson inequality

- $\Omega \stackrel{\text{def}}{=} 2^{[n]}$, $0 \leq p_i \leq 1$ for $i=1, 2, \dots, n$.
- $(\Omega, 2^\Omega, P)$ is a prob. space s.t. $P(\{\omega\}) = \prod_{i \in \omega} p_i \prod_{j \in \bar{\omega}} (1-p_j)$ for $\forall \omega \in \Omega$.
- Let $A_1, \dots, A_m \subseteq \Omega$ be fixed s.t. $\underset{\text{ordered pair}}{i \sim j} \iff A_i \cap A_j \neq \emptyset$.
- $B_i \stackrel{\text{def}}{=} \{\omega \in \Omega : A_i \subseteq \omega\}$, $i=1, 2, \dots, m$.

Thm Suppose $P(B_i) \leq \varepsilon < 1$, $\forall i$. $\mu \stackrel{\text{def}}{=} \sum_{i \in [m]} P(B_i)$, $M \stackrel{\text{def}}{=} \prod_{i \in [m]} P(\bar{B}_i)$ and $\Delta \stackrel{\text{def}}{=} \sum_{i \sim j} P(B_i \cap B_j)$. Then we have

$$(1) M \stackrel{\textcircled{1}}{\leq} P\left(\bigcap_{i=1}^m \bar{B}_i\right) \stackrel{\textcircled{2}}{\leq} M e^{\frac{\Delta}{2(1-\varepsilon)}}, \text{ and}$$

$$(2) P\left(\bigcap_{i=1}^m \bar{B}_i\right) \stackrel{\textcircled{3}}{\leq} e^{-\mu + \frac{\Delta}{2}}$$

pf

• claimA For $i, k \in J \subseteq [m]$, we have

$$\textcircled{a} \quad P(\bar{B}_i | \bigwedge_{j \in J} \bar{B}_j) \geq P(\bar{B}_i), \quad \text{and}$$

$$\textcircled{b} \quad P(B_i | B_k \cap \bigwedge_{j \in J} \bar{B}_j) \leq P(B_i | B_k).$$

pf(claimA): $\omega' \subseteq \omega \in \bar{B}_i \Rightarrow \omega' \in \bar{B}_i$ and $\omega' \supseteq \omega \in B_i \Rightarrow \omega' \in B_i$.

\textcircled{a} \bar{B}_i and $\bigcap_{j \in J} \bar{B}_j$ are downsets $\Rightarrow P(\bar{B}_i \cap \beta) \geq P(\bar{B}_i) P(\beta) (\because H\text{-ineq.})$

\textcircled{b} B_i is an upset and β is a downset $\Rightarrow P_{P_1=P_2=\dots=P_\ell=1}(B_i \cap \beta) \leq P_{P_1=P_2=\dots=P_\ell=1}(B_i) P_{P_1=P_2=\dots=P_\ell=1}(\beta) (\because H\text{-ineq.})$

(where we assume $B_K = \{1, 2, \dots, \ell\}$ w.l.o.g) $\Rightarrow P(B_i \cap \beta | B_K) \leq P(B_i | B_K) P(\beta | B_K)$

$$\Rightarrow \frac{P(B_i \cap \beta \cap B_K)}{P(\beta \cap B_K)} \leq P(B_i | B_K) \Rightarrow P(B_i | B_K \cap \beta) \leq P(B_i | B_K) \text{ done!}$$

QED of claimA

$$\underline{Pf(\text{ineq } \textcircled{1})} \quad P\left(\bigcap_{i=1}^m \bar{B}_i\right) = \prod_{i=1}^m P(\bar{B}_i | \bigcap_{1 \leq j < i} \bar{B}_j) \quad (\text{let } \bigcap_{1 \leq j < i} \bar{B}_j = \Omega)$$

$$\geq \prod_{i=1}^m P(\bar{B}_i) \quad (\because \text{claimA } \textcircled{a})$$

pf (continued) Pf (ineq ②) $P(\bigcap_{i=1}^m \bar{B}_i) = \prod_{i=1}^m P(\bar{B}_i | \bigcap_{1 \leq j < i} \bar{B}_j)$

$$= \prod_{i=1}^m \left\{ 1 - P(B_i \cap \bigwedge_{j \sim i, j < i} \bar{B}_j \cap \bigwedge_{j > i} \bar{B}_j) / P(B \cap C) \right\} \leq \prod_{i=1}^m \left\{ 1 - \frac{P(B_i \cap B \cap C)}{P(B_i \cap C)} \frac{P(B_i \cap C)}{P(C)} \right\}$$

$$= \prod_{i=1}^m \left\{ 1 - P(B_i | B_i \cap C) P(B_i) \right\} \leq \prod_{i=1}^m \left\{ 1 - \left\{ 1 - \sum_{j \sim i, j < i} P(B_j | B_i \cap C) \right\} P(B_i) \right\}$$

$\because P(B_i \cap C) = P(B_i) P(C)$

$$\leq \prod_{i=1}^m \left\{ 1 - \left\{ 1 - \sum_{j \sim i, j < i} P(B_j | B_i) \right\} P(B_i) \right\} = \prod_{i=1}^m \left\{ 1 - P(B_i) + \sum_{j \sim i, j < i} P(B_i \cap B_j) \right\} = \star$$

$$\leq \prod_{i=1}^m P(\bar{B}_i) \left\{ 1 + \frac{1}{1-\varepsilon} \underbrace{\sum_{j \sim i, j < i} P(B_i \cap B_j)}_{\alpha_i} \right\} \leq \prod_{i=1}^m P(\bar{B}_i) e^{\alpha_i} = M e^{\frac{1}{1-\varepsilon} \sum_{i=1}^m \alpha_i} = M e^{\frac{\Delta}{2(1-\varepsilon)}}.$$

Pf (ineq ③) $P(\bigcap_{i=1}^m \bar{B}_i) \leq \star$

$$\leq \prod_{i=1}^m \exp \left\{ -P(B_i) + \alpha_i \right\}$$

$$= \exp \left\{ -\mu + \sum_{i=1}^m \alpha_i \right\} = e^{-\mu + \frac{\Delta}{2}}$$

QED

The XYZ-Theorem (I)

Thm

(shepp 1982) Let the incomes x_1, \dots, x_n of n

individuals be initially ordered at random uniformly
on all permutations. Suppose some partial information
is available on the true ordering of the x 's;

e.g. $\Gamma = \{x_1 < x_{12}, x_1 < x_5, \dots\}$. Then for any Γ ,
we have

$$P(x_1 < x_2 | \Gamma) \leq P(x_1 < x_2 | \Gamma, x_1 < x_3).$$

Pf: $[N]^n \stackrel{\text{def}}{=} \overbrace{[N] \times \cdots \times [N]}^{n \text{ terms}}$

If $x, y \in [N]^n$ then we write $x = (x_1, \dots, x_n)$ and $y = (y_1, \dots, y_n)$.

Claim a. $([N]^n, \leq)$ is a lattice, where $x \leq y$ iff

$x_i \geq y_i$ and $x_i - x_1 \leq y_i - y_1$ for $i = 1, 2, 3, \dots, n$.

Pf of claim a: Clearly $([N]^n, \leq)$ is a poset and that

$$(x \wedge y)_i = \min(x_i - x_1, y_i - y_1) + \max(x_1, y_1), \quad i = 1, 2, \dots, n$$

$$(x \vee y)_i = \max(x_i - x_1, y_i - y_1) + \min(x_1, y_1), \quad i = 1, 2, \dots, n$$

Thus $([N]^n, \leq)$ is a lattice.

Note: we need to check that $(x \wedge y)_i, (x \vee y)_i \in [N]$! Prove it!

Exercise

END

Pf (continued) Claim b. $([N]^n, \leq)$ is a distributive lattice.

pf of claim b: $(x \wedge (y \vee z))_i$

$$= \min(x_i - x_1, (y \vee z)_i - (y \vee z)_1) + \max(x_1, (y \vee z)_1)$$

$$= \min(x_i - x_1, \max(y_i - y_1, z_i - z_1)) + \max(x_1, \min(y_1 - z_1))$$

$$= \max(\min(x_i - x_1, y_i - y_1), \min(x_i - x_1, z_i - z_1)) + \min(\max(x_1, y_1), \max(x_1, z_1))$$

$$= \max((x \wedge y)_i - (x \wedge y)_1, (x \wedge z)_i - (x \wedge z)_1) + \min((x \wedge y)_1, (x \wedge z)_1)$$

$$= ((x \wedge y) \vee (x \wedge z))_i$$

END

pf (continued) Claim c let $f(x) \stackrel{\text{def}}{=} \begin{cases} 1 & \text{if } x_1 \leq x_2 \\ 0 & \text{o.w.} \end{cases}$ $g(x) \stackrel{\text{def}}{=} \begin{cases} 1 & \text{if } x_1 \leq x_3 \\ 0 & \text{o.w.} \end{cases}$

then $f, g: [N]^n \rightarrow \{0,1\}$ are two increasing functions.

pf of claimc: $x \leq y$ and $f(x)=1 \Rightarrow x \leq y$ and $x_1 \leq x_2 \Rightarrow x_2 - x_1 \leq y_2 - y_1$, and $x_1 \leq x_2$.
Therefore we arrive at $y_1 \leq y_2$ and hence $f(y)=1$.

END

Claim d. Let $\mu: [N]^n \rightarrow \mathbb{R}_{\geq 0}$ s.t. $\mu(x) = \begin{cases} 1 & \text{if } x \text{ satisfies the inequalities in } \Gamma \\ 0 & \text{o.w.} \end{cases}$

Then μ is log-supermodular.

pf of claimd: For $z, y \in [N]^n$ if $\mu(z)=\mu(y)=1$, and $x_i < x_j$ is one of the inequalities in Γ . Then we have $z_i < z_j$ & $y_i < y_j$.

$$\begin{aligned} (z \wedge y)_i &= \min(z_i - z_1, y_i - y_1) + \max(z_1, y_1) \\ &\leq \min(z_j - z_1, y_j - y_1) + \max(z_1, y_1) = (z \wedge y)_j \end{aligned}$$

Similarly, $(z \vee y)_i \leq (z \vee y)_j$. Thus $\mu(z \wedge y) = \mu(z \vee y) = 1$.
Therefore $\mu(z)\mu(y) \leq \mu(z \wedge y)\mu(z \vee y)$.

END

Pf

(continued)

Claims a, b, c, d + FKG inequality

$$\Rightarrow \left(\sum_{x \in [N]^n} \mu(x) f(x) \right) \left(\sum_{x \in [N]^n} \mu(x) g(x) \right) \leq \left(\sum_{x \in [N]^n} \mu(x) f(x) g(x) \right) \left(\sum_{x \in [N]^n} \mu(x) \right)$$

$$\Rightarrow P(x_1 \leq x_2, \Gamma) P(x_1 \leq x_3, \Gamma) \leq P(x_1 \leq x_2, x_1 \leq x_3, \Gamma) P(\Gamma)$$

\Rightarrow As $N \rightarrow \infty$ we note that $P(x_i = x_j) \rightarrow 0$ provided $i \neq j$, and hence

$$P(x_1 < x_2, \Gamma) P(x_1 < x_3, \Gamma) \leq P(x_1 < x_2, x_1 < x_3, \Gamma) P(\Gamma)$$

$$\Rightarrow P(x_1 < x_2 | \Gamma) \leq P(x_1 < x_2 | x_1 < x_3, \Gamma)$$

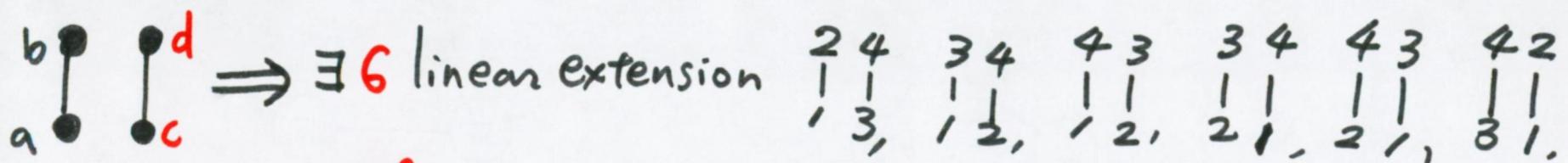
QED

Remark: You might note that in the above proof we choose lattice $([N]^n, \leq)$ first, and then let $N \rightarrow \infty$ to get what we want. Can you explain why don't we replace $([N]^n, \leq)$ by (S_n, \leq) ? $S_n \leftarrow$ the set of permutation on integers $1, 2, \dots, n$

Linear Extension

- Let (P, \leq) be a poset with n elements.
 A mapping $\sigma: P \xrightarrow{\text{H}} [n]$ is called a linear extension of P
 if $x \leq y \Rightarrow \sigma(x) \leq \sigma(y)$
- In the remaining slides, we consider the probability space of
 all linear extension of P , where each sample point is equally likely.

Example: Consider a Hasse diagram of 4-elements as follows.



$\begin{smallmatrix} 2 & 4 \\ & \\ 1 & 3 \end{smallmatrix}$	$\begin{smallmatrix} 3 & 4 \\ & \\ 1 & 2 \end{smallmatrix}$	$\begin{smallmatrix} 4 & 3 \\ & \\ 1 & 2 \end{smallmatrix}$	$\begin{smallmatrix} 3 & 4 \\ & \\ 2 & 1 \end{smallmatrix}$	$\begin{smallmatrix} 4 & 3 \\ & \\ 2 & 1 \end{smallmatrix}$	$\begin{smallmatrix} 4 & 2 \\ & \\ 3 & 1 \end{smallmatrix}$
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$P(x \leq y) \stackrel{\text{def}}{=} \text{the probability that } \sigma(x) \leq \sigma(y) \text{ in a randomly chosen linear extension } \sigma.$

For example, $P(c \leq b) = P(\{\sigma : \sigma(c) \leq \sigma(b)\}) = \frac{5}{6}$, $P(c \leq d) = \frac{6}{6}$.

The XYZ - Theorem (II)

Thm 6.4.1^{p88} Let P be a poset with n elements a_1, a_2, \dots, a_n . Then

$$P(a_1 \leq a_2 \text{ and } a_1 \leq a_3) \leq P(a_1 \leq a_2) P(a_1 \leq a_3)$$

A false conjecture: Is $P(x_1 < x_2 < x_4 | \Gamma) \leq P(x_1 < x_2 < x_4 | \Gamma, x_1 < x_3 < x_4)$?

Sol: In general its not true!

$$\text{Let } n=6 \text{ and } \Gamma = \{x_2 < x_5 < x_6 < x_3, x_1 < x_4\}$$

$$\text{Then we have } P(x_1 < x_2 < x_4 | \Gamma) = \frac{4}{15}.$$

$$\text{However } P(x_1 < x_2 < x_4 | \Gamma, x_1 < x_3 < x_4) = \frac{1}{4}.$$

END