

# Coupling & FKG.

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5:57 PM

Today: Stochastic domination, Strassen's theorem

Correlation inequalities: FKG; proof via Glauber dynamics

- Coupling: Let  $\mu, \nu$  be prob measures on  $\Omega$ ; a coupling of  $\mu, \nu$  is a probability measure  $\gamma$  on  $\Omega \times \Omega$

$$\text{s.t. } \gamma(A \times \Omega) = \mu(A), \quad \gamma(\Omega \times A) = \nu(A), \quad \forall A \subseteq \Omega \text{ measurable}$$

Example: Poissonization of independent Bernoulli

$X_1 \sim \text{Ber}(p_1)$ ,  $X_2 \sim \text{Ber}(p_2)$ , ...,  $X_n \sim \text{Ber}(p_n)$ , independently

$$S = X_1 + \dots + X_n$$

Then  $S$  can be coupled to  $Z \sim \text{Poi}(\lambda)$ ,  $\lambda = \lambda_1 + \dots + \lambda_n$ ,  $\lambda_i = -\log(1-p_i)$

$$\text{s.t. } \mathbb{P}[S \neq Z] \leq \frac{1}{2} \sum_{i=1}^n \lambda_i^2 \quad (\rightarrow 0 \text{ if } f \propto \lambda, \max_{i \in [n]} \lambda_i \rightarrow 0)$$

Proof:  $W_i = \text{Poi}(\lambda_i)$ ;  $Z = \sum_{i=1}^n W_i$

$$e^{-\lambda_i} = 1 - p_i = \mathbb{P}[W_i = 0] = \mathbb{P}[X_i = 0]$$

$$\Rightarrow \text{can couple } W_i, X_i \text{ s.t. } \mathbb{P}[W_i \neq X_i] = \mathbb{P}[W_i \geq 1] \leq \frac{\lambda_i}{2}$$

- Stochastic domination:

Real variables:  $X$  stochastically dominates  $Y$ , if  $\mathbb{P}[X > a] \geq \mathbb{P}[Y > a]$ , for any  $a \in \mathbb{R}$

Example:  $\text{Po}(\lambda)$  s.d.  $\text{Ber}(p)$ ,  $\lambda \geq -\log(1-p)$

Then  $X$  s.d.  $Y$ , iff there is a coupling  $(\tilde{X}, \tilde{Y})$  of  $X, Y$ , s.t.  $\mathbb{P}[\tilde{X} \geq \tilde{Y}] = 1$

Point: If such a coupling exists, for any  $a \in \mathbb{R}$

$$\mathbb{P}[Y > a] = \mathbb{P}[\tilde{Y} > a] = \mathbb{P}[\tilde{X} \geq \tilde{Y} > a] \leq \mathbb{P}[\tilde{X} > a] = \mathbb{P}[X > a]$$

$$\textcircled{2} \text{ If } X \text{ s.d. } Y: \text{ define } f_X(a) = \mathbb{P}[X \leq a], f_Y(a) = \mathbb{P}[Y \leq a] \Rightarrow f_X(a) \leq f_Y(a)$$

Take  $U \sim \text{unif}[\mathbb{0}, 1]$ . Let  $\tilde{X} = f_X^{-1}(U)$ ,  $\tilde{Y} = f_Y^{-1}(U)$

$$\tilde{X} \stackrel{d}{=} X, \quad \tilde{Y} \stackrel{d}{=} Y, \quad \mathbb{P}[\tilde{X} \geq \tilde{Y}] = \mathbb{P}[f_X^{-1}(U) \geq f_Y^{-1}(U)] = 1$$

Cor. For  $X$  s.d.  $Y$ , and  $f: \mathbb{R} \rightarrow \mathbb{R}$  non-decreasing,  $f(X)$  s.d.  $f(Y)$

And if  $\mathbb{E}|f(x)|, \mathbb{E}|f(y)| < \infty$ ,  $\mathbb{E}f(x) \geq \mathbb{E}f(y)$

## Partially ordered sets, (POSET)

$(\Omega, \leq)$ : for any  $x, y, z \in \Omega$

$$\textcircled{1} \quad x \leq x$$

$$\textcircled{2} \quad \text{if } x \leq y, y \leq z \Rightarrow x \leq z$$

$$\textcircled{3} \quad \text{if } x \leq y, y \leq z \Rightarrow x \leq z$$

Example:  $\mathbb{R}^d$ :  $(x_1, \dots, x_d) \leq (y_1, \dots, y_d)$  iff  $x_i \leq y_i$  for each  $i \in \{1, \dots, d\}$

Increasing set:  $A \subseteq \Omega$  s.t.  $x \in A, x \leq y \Rightarrow y \in A$

Increasing function:  $f: \Omega \rightarrow \mathbb{R}$  s.t.  $x \leq y \Rightarrow f(x) \leq f(y)$

Stochastic domination:  $X, Y$  random in  $\Omega$ ,  $X$  s.d.  $Y$  if  $\mathbb{P}[X \in A] \geq \mathbb{P}[Y \in A] \forall$  increasing  $A$

(Strassen's thm)  $X, Y$  be random variables in a finite poset  $(\Omega, \leq)$

$X$  s.d.  $Y$  iff there is a coupling  $(\tilde{X}, \tilde{Y})$  s.t.  $\tilde{X} \geq \tilde{Y}$

Proof:  $\textcircled{1}$  If  $\exists$  coupling,  $\mathbb{P}[Y \in A] \leq \mathbb{P}[X \in A] \forall$  increasing  $A$

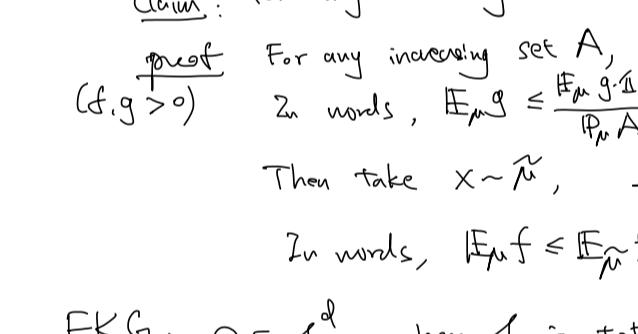
$\textcircled{2}$  If  $X$  s.d.  $Y$ , construct coupling.

i.e. give  $\mu_X, \mu_Y$  on  $\Omega$ , want  $\nu: \Omega \times \Omega \rightarrow \mathbb{R}_{\geq 0}$

s.t.  $\nu(x, y) > 0$  only for  $x \geq y$ ; and  $\sum_y \nu(x, y) = \mu_X(x)$ ,  $\sum_x \nu(x, y) = \mu_Y(y)$

Idea: max-flow = min-cut theorem

$$\Omega = \{x, y, z, w\}$$



$$x_1 \rightarrow y_1 \text{ iff } x \geq y, \text{ cap}(x_1 \rightarrow y_1) = \infty$$

$$\text{cap}(r \rightarrow x_1) = \mu_X(x_1), \quad \text{cap}(x_2 \rightarrow s) = \mu_Y(x_2)$$

$$\text{If max flow} = 1 \Rightarrow \nu(x, y) = \text{flow}(x \rightarrow y)$$

Theorem: max flow = min cut

$$R: \{x : (x_2 \rightarrow s) \text{ is cut}\} \quad L: \{x : (r \rightarrow x_1) \text{ is cut}\} \supseteq L^*, \{x : x \geq y, \exists y \in R\}$$

$L^*$  is increasing

$L^* \supseteq \Omega \setminus R$

$$\text{cutsize} = \mu_X(L) + \mu_Y(R) \geq \mu_X(L^*) + \mu_Y(\Omega \setminus L^*) = \mu_X(L^*) + \mu_Y(\Omega \setminus L^*) \geq 1.$$

Cor.  $X, Y$  random variables in  $\Omega$ ,  $X$  s.d.  $Y \Rightarrow f(x)$  s.d.  $f(y)$ ; if  $|\mathbb{E}f(x)|, |\mathbb{E}f(y)| < \infty$ ,  $\mathbb{E}f(x) \geq \mathbb{E}f(y)$

$f: \Omega \rightarrow \mathbb{R}$  increasing function

(Positive association) A measure  $\mu$  on poset  $\Omega$  has P.a. if for any two increasing events  $A, B$ ,  $\mu(A \cap B) \geq \mu(A)\mu(B)$

Claim: for any increasing functions  $f, g: \Omega \rightarrow \mathbb{R}$ , with  $\mathbb{E}|f|, \mathbb{E}|g| < \infty$ ,  $\mathbb{E}fg \geq \mathbb{E}f\mathbb{E}g$

Proof: For any increasing set  $A$ , take  $X \sim \frac{\mu_A}{\mu(A)}$ ,  $Y \sim \mu$ ; then  $X$  s.d.  $Y$ , and in particular  $\mathbb{E}g(X) \geq \mathbb{E}g(Y)$

( $f, g > 0$ ) In words,  $\mathbb{E}fg \leq \frac{\mu_A f \mu_B g}{\mu(A)\mu(B)}$

Then take  $X \sim \tilde{\mu}$ ,  $\frac{\tilde{\mu}}{\mu} = \frac{1}{\mathbb{E}f}$ ,  $Y \sim \mu \Rightarrow \tilde{\mu}(A) \geq \mu(A)$ ; then  $X$  s.d.  $Y$ , and in particular  $\mathbb{E}f(X) \geq \mathbb{E}f(Y)$

In words,  $\mathbb{E}_\mu f \leq \frac{\mathbb{E}_\mu fg}{\mathbb{E}_\mu g}$

FKG:  $\Omega = \mathbb{A}^d$ , when  $\mathbb{A}$  is totally ordered (then  $\Omega$  poset)

If a measure  $\mu$  satisfies  $\mu(w \vee w') \mu(w \wedge w') \geq \mu(w) \mu(w')$  CFKG condition

$$\text{then } \mu \text{ has pos. ass.}$$

$$\max \quad \text{out-} \quad \min \quad \text{out-} \quad \text{wise}$$

$$\text{e.g. } \mu \text{ is a product measure. } \mu(w \vee w') \mu(w \wedge w') = \prod_{i=1}^d \mu_i(w_i \vee w'_i) \mu_i(w_i \wedge w'_i) = \prod_{i=1}^d \mu_i(w_i) \mu_i(w'_i) = \mu(w) \mu(w')$$

example: In Erdős-Rényi graph  $G(n, p)$

A:  $G(n, p)$  is connected

B: chromatic number  $\geq 4$

$$\mathbb{P}[A \wedge B] \geq \mathbb{P}[A] \mathbb{P}[B]$$

Proof of FKG (assume  $\Omega$  finite, for correctness)

Take  $A \subseteq \Omega$  an increasing set;  $X \sim \frac{\mu_A}{\mu(A)}$ ,  $Y \sim \mu$

suffices to show that  $X$  s.d.  $Y$ .

Idea: coupled Markov chain

Glauber dynamics for  $\tilde{X} = (\tilde{x}_1, \dots, \tilde{x}_d)$ ,  $\tilde{Y} = (\tilde{y}_1, \dots, \tilde{y}_d)$

choose  $i \in \{1, \dots, d\}$  uniformly random

replace  $\tilde{x}_i$  by  $\tilde{x}'_i \sim \frac{\mu_A}{\mu(A)} (\cdot | \tilde{x}_1, \dots, \tilde{x}_{i-1}, \tilde{x}_{i+1}, \dots, \tilde{x}_d)$

$\tilde{y}_i$  by  $\tilde{y}'_i \sim \mu (\cdot | \tilde{y}_1, \dots, \tilde{y}_{i-1}, \tilde{y}_{i+1}, \dots, \tilde{y}_d)$

Law of  $\tilde{X} \rightarrow \frac{\mu_A}{\mu(A)}$  since they are the unique stationary measures of these Markov chains

Law of  $\tilde{Y} \rightarrow \mu$

Let  $\tilde{X} \geq \tilde{Y}$  initially, can ensure  $\tilde{X} \geq \tilde{Y}$  after each step.

Indeed, by FKG condition, for  $x_1 \geq y_1, \dots, x_n \geq y_n$ ,  $x_i \geq y_i, \dots, x_n \geq y_n$  (since for any  $a \geq b$ ,

$$\mu(\cdot | x_1, \dots, x_n, a, y_1, \dots, y_n) \geq \frac{\mu(x_1, \dots, x_n, a) \mu(y_1, \dots, y_n)}{\mu(x_1, \dots, x_n, b, y_1, \dots, y_n)} \geq \frac{\mu(y_1, \dots, y_n, a)}{\mu(y_1, \dots, y_n, b)}$$

$$\text{Also } \frac{\mu_A}{\mu(A)} (\cdot | x_1, \dots, x_n, y_1, \dots, y_n) \text{ s.d. } \mu(\cdot | x_1, \dots, x_n, y_1, \dots, y_n)$$

$$\Rightarrow \tilde{X} \geq \tilde{Y} \text{ at time } \infty, \quad X \text{ s.d. } Y.$$

Another example: Ising model (ferromagnetic)

In a finite graph  $G$ ,  $\{6_v\}_{v \in V} \in \{\pm 1\}^V$

measure  $\mu$  on  $\{\pm 1\}^V$

$$\text{s.t. } \mu(\{6_v\}_{v \in V}) = \frac{1}{Z} \exp(-\beta \sum_{u \sim v} 6_u 6_v)$$

$$Z = \sum_{\{6_v\}_{v \in V}} \exp(-\beta \sum_{u \sim v} 6_u 6_v)$$

$$= \prod_{v \in V} \sum_{6_v \in \{-1, 1\}} \exp(-\beta \sum_{u \sim v} 6_u 6_v)$$

$$= 2^{N_G} \prod_{v \in V} 2$$

$$\geq \mu(w) \mu(w')$$

$$\Rightarrow \mathbb{E}[6_u 6_v] \geq \mathbb{E}[6_u] \mathbb{E}[6_v]$$

by symmetry

Example:  $(X_{ij})_{i,j=1}^n$  be  $n \times n$  matrix with independent random entries

$\|X\|_2$  operator norm &  $\text{tr}X$  positively correlated.

$$(\mathbb{P}[\|X\|_2 > a, \text{tr}X > b] \geq \mathbb{P}[\|X\|_2 > a] \mathbb{P}[\text{tr}X > b])$$

First-passage percolation:  $T(u, v) = \min_{\gamma \in \Gamma} \sum_{e \in \gamma} w_e$   $\{w_e\}_{e \in E}$  independent

$T(u, v), T(u', v')$  positively correlated

$$(\mathbb{P}[T(u, v) > a, T(u', v') > b] \geq \mathbb{P}[T(u, v) > a] \mathbb{P}[T(u',$$