# Topic Course on Probabilistic Methods (Week 6) <br> Correlation Inequalities 

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## Introduction

The topic course is mostly based the textbook "The probabilistic Method" by Noga Alon and Joel Spencer (third edition 2008, John Wiley \& Sons, Inc. ISBN 9780470170205 or fourth edition ISBN-13: 978-1119061953.)


## Selected topics

■ Linearity of Expectation (2 weeks)

- Alterations (1 week)
- The second moment method (1 week)
- The Local Lemma (1-2 weeks)
- Correlation Inequalities (1 week)
- Large deviations (1-2 weeks)
- Poisson Paradigm (1 week)
- Random graphs (2 weeks)
- Discrepancy (1 week)
- Entropy (1 week)


## Subtopics

## Correlation Inequalities

- Four Functions Theorem
- 4FT on distributive lattice
- FKG inequalities


## Correlation Inequalities

- $(\Omega, \mathcal{F}, P)$ : a probability space.
$A, B$ : two events.
$A$ and $B$ are independent if

$$
\operatorname{Pr}(A B)=\operatorname{Pr}(A) \operatorname{Pr}(B)
$$

$A$ and $B$ are positively correlated if

$$
\operatorname{Pr}(A B) \geq \operatorname{Pr}(A) \operatorname{Pr}(B)
$$

$A$ and $B$ are negatively correlated if

$$
\operatorname{Pr}(A B) \leq \operatorname{Pr}(A) \operatorname{Pr}(B)
$$

## Four Functions Theorem

- $N:=\{1,2,3 \ldots, n\}$
- $P(N)$ : the power set of $N$.
- $\alpha, \beta, \gamma, \delta: P(N) \rightarrow \mathbb{R}^{+}$
- For $\mathcal{A} \subset P(N)$, and $\phi \in\{\alpha, \beta, \gamma, \delta\}$, let $\phi(\mathcal{A})=\sum_{A \in \mathcal{A}} \phi(A)$.


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$$
\phi(\mathcal{A})=\sum_{A \in \mathcal{A}} \phi(A)
$$

Theorem [Ahlswede, Daykin (1978)]: If for any $A, B \subset N$,

$$
\alpha(A) \beta(B) \leq \gamma(A \cup B) \delta(A \cap B)
$$

then for any $\mathcal{A}, \mathcal{B} \subset P(N)$,

$$
\alpha(\mathcal{A}) \beta(\mathcal{B}) \leq \gamma(\mathcal{A} \cup \mathcal{B}) \delta(\mathcal{A} \cap \mathcal{B})
$$

## Proof

## Simplification:

- Modifying $\alpha$ so that $\alpha(A)=0$ for all $A \notin \mathcal{A}$.
- Modifying $\beta$ so that $\beta(B)=0$ for all $B \notin \mathcal{B}$.

■ Modifying $\gamma$ so that $\gamma(C)=0$ for all $C \notin \mathcal{A} \cup \mathcal{B}$.

- Modifying $\delta$ so that $\delta(D)=0$ for all $D \notin \mathcal{A} \cap \mathcal{B}$.


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- Modifying $\gamma$ so that $\gamma(C)=0$ for all $C \notin \mathcal{A} \cup \mathcal{B}$.
- Modifying $\delta$ so that $\delta(D)=0$ for all $D \notin \mathcal{A} \cap \mathcal{B}$.

$$
\alpha(A) \alpha(B) \leq \gamma(A \cup B) \delta(A \cap B)
$$

still holds. It is sufficient to prove for $\mathcal{A}=\mathcal{B}=P(N)$.

## Induction on $n$

Initial case $n=1: P(N)=\{\emptyset, N\}$. Use index 0 for $\emptyset$ and 1 for $N$. We have

$$
\begin{aligned}
& \alpha_{0} \beta_{0} \leq \gamma_{0} \delta_{0} \\
& \alpha_{0} \beta_{1} \leq \gamma_{1} \delta_{0} \\
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& \alpha_{1} \beta_{1} \leq \gamma_{1} \delta_{1} .
\end{aligned}
$$

We need prove

$$
\left(\alpha_{0}+\alpha_{1}\right)\left(\beta_{0}+\beta_{1}\right) \leq\left(\gamma_{0}+\gamma_{1}\right)\left(\delta_{0}+\delta_{1}\right)
$$

It can be directly verified.

## Inductive step

Suppose it holds for $n-1$ and let us prove it for $n \geq 2$. Let $N^{\prime}=N \backslash\{n\}$ and for each $\phi \in\{\alpha, \beta, \gamma, \delta\}$ and $A \in N^{\prime}$ define

$$
\phi^{\prime}(A)=\phi(A)+\phi(A \cup\{n\}) .
$$

Note that $\phi(P(N))=\phi^{\prime}\left(P\left(N^{\prime}\right)\right)$. Apply inductive hypothesis for functions $\alpha^{\prime}, \beta^{\prime}, \gamma^{\prime}$, and $\delta^{\prime}$. It suffices to check

$$
\alpha^{\prime}(A) \alpha^{\prime}(B) \leq \gamma^{\prime}(A \cup B) \delta^{\prime}(A \cap B) .
$$

This is similar to the case $n=1$.

## Distributive lattice

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$$
a \wedge(b \wedge c)=(a \wedge b) \wedge c
$$

- Absorption laws: $a \vee(a \wedge b)=a, a \wedge(a \vee b)=a$.

It is distributive if it further satisfies the distributive laws:

$$
\begin{aligned}
& a \wedge(b \vee c)=(a \wedge b) \vee(a \wedge c), \\
& a \vee(b \wedge c)=(a \vee b) \wedge(a \vee c) .
\end{aligned}
$$

## 4FT on distributive lattice

Theorem [Ahlswede, Daykin (1978)]: Let $L$ be a distributive lattice and $\alpha, \beta, \gamma, \delta: L \rightarrow \mathbb{R}^{+}$. If for any $x, y \in L$,

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\alpha(x) \alpha(y) \leq \gamma(x \vee y) \delta(x \wedge y)
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then for any $X, Y \subset L$,

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Note any distributive lattice can be embedded into $P([n])$. This is a corollary of the previous theorem.

## FKG inequalities

A function $\mu: L \rightarrow \mathbb{R}^{+}$is log-supermodular if $\mu(x) \mu(y) \leq \mu(x \vee y) \mu(x \wedge y)$ for all $x, y$.

## FKG inequalities

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■ $f: L \rightarrow \mathbb{R}^{+}$is increasing if $f(x) \leq f(y)$ whenever $x \leq y$. It is decreasing if $f(x) \geq f(y)$ whenever $x \leq y$.

The FKG Inequality [Fortuin-Kasteleyn-Ginibre 1971]: If $\mu$ is log-supermodular and $f, g$ are increasing, then
$\sum_{x \in L} f(x) \mu(x) \sum_{x \in L} g(x) \mu(x) \leq \sum_{x \in L} f(x) g(x) \mu(x) \sum_{x \in L} \mu(x)$.

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$$

If one is increasing and the other is decreasing, then

$$
\sum_{x \in L} f(x) \mu(x) \sum_{x \in L} g(x) \mu(x) \geq \sum_{x \in L} f(x) g(x) \mu(x) \sum_{x \in L} \mu(x) .
$$

## A probabilistic view

■ $\quad(P(N), \mu)$ : a probability space where $\mu$ is log-supermodular.

- An event $\mathcal{A}$ is monotone increasing if $A \in \mathcal{A}$ and $A \subset B$ implies $B \in \mathcal{A}$.

Proposition: If both $A$ and $B$ are monotone increasing or monotone decreasing, then

$$
\operatorname{Pr}(\mathcal{A B}) \geq \operatorname{Pr}(\mathcal{A}) \operatorname{Pr}(\mathcal{B})
$$

If one is monotone increasing and the other one is monotone decreasing, then

$$
\operatorname{Pr}(\mathcal{A B}) \leq \operatorname{Pr}(\mathcal{A}) \operatorname{Pr}(\mathcal{B})
$$

## Applying to $G(n, p)$

In $G(n, p)$, for any graph $H$,

$$
\mu(H)=\operatorname{Pr}(H)=p^{|E(H)|}(1-p)^{|E(\bar{H})|} .
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Observe that this $\mu$ is log-supermodular. We get a lot of correlation inequalities on monotone events.

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Example of monotone events:

- Triangle-free.
- Being planar graph.
- $k$-connected.
- Having Hamiltonian cycle.

■ $H$-free.
Diameter less than $k$.

