## The Finite-Dimensional

## Witsenhausen Counterexample

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Joint work with Prof. Anant Sahai, Se Yong Park
There are handouts for this talk, please take one!

Outline

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- Witsenhausen's counterexample


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- Witsenhausen's counterexample
- Infinite-length vector extension


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## It is easier to approximate

## Witsenhausen's counterexample

$$
\begin{aligned}
& x_{0} \sim \mathcal{N}\left(0, \sigma_{0}^{2}\right)
\end{aligned}
$$

$$
\begin{aligned}
& w \sim \mathcal{N}(0,1) \quad \min \left\{k^{2} \mathbb{E}\left[u_{1}^{2}\right]+\mathbb{E}\left[x_{2}^{2}\right]\right\}
\end{aligned}
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"Implicit channel"


$$
\mathbb{E}\left[u_{1}^{2}\right] \leq P
$$

$\min \mathbb{E}\left[\left(x_{1}-\hat{x}_{1}\right)^{2}\right]$

## Implicit channel based signaling strategies

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[Mitter and Sahai, '99]

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## Infinite-length counterexample

## A simplification


[Ho, Kastner, Wong '78]

## Vector quantization



## Vector quantization



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## Vector quantization



## A lower bound for the vector extension

$$
\begin{aligned}
& \overline{\mathcal{C}}_{\text {min }} \geq \inf _{P \geq 0} k^{2} P+\left((\sqrt{\kappa(P)}-\sqrt{P})^{+}\right)^{2} \\
& \kappa(P)=\frac{\sigma_{0}^{2}}{\left(\sigma_{0}+\sqrt{P}\right)^{2}+1}
\end{aligned}
$$

## Optimal costs within a factor of 4.45



## A Dirty Paper Coding-based strategy

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## A Dirty Paper Coding-based strategy



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## A Dirty Paper Coding-based strategy



[Baglietto, Parisini, Zoppoli]
[Lee, Lau and Ho]

## Optimal cost to within a factor of 2



The scalar problem

## Quantization-based strategies



## Vector lower bound is too loose at finite lengths!



## Another look at the vector lower bound

$$
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& \overline{\mathcal{C}}_{\min } \geq \inf _{P \geq 0} k^{2} P+\left((\sqrt{\kappa(P)}-\sqrt{P})^{+}\right)^{2} \\
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## Sphere-packing extension of lower bound



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## Scalar case : Quantization based strategies are approximately optimal!



Finite vector lengths

## "Good" strategies



## "Good" strategies



## "Good" strategies



## What makes "good" lattices?



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What makes "good" lattices?


What makes "good" lattices?


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## What makes "good" lattices?



## What makes "good" lattices?



## 2-D case



## Lattices are uniformly approximately optimal over dimension size

$$
\begin{aligned}
& \inf _{P \geq 0} k^{2} P+\eta\left(P, \sigma_{0}^{2}\right) \leq \bar{J} \leq \mu\left(\inf _{P \geq 0} k^{2} P+\eta\left(P, \sigma_{0}^{2}\right)\right) \\
& \mu \leq 300 \zeta^{2}, \quad \zeta \leq 4
\end{aligned}
$$

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Papers/slides/handouts available at : http://www.eecs.berkeley.edu/~pulkit/

## Back-up slides begin

