

On entropy-constrained Gaussian channel capacity via the moment problem

Adway Girish

joint work with Shlomo Shamai and Emre Telatar



June 26, 2025

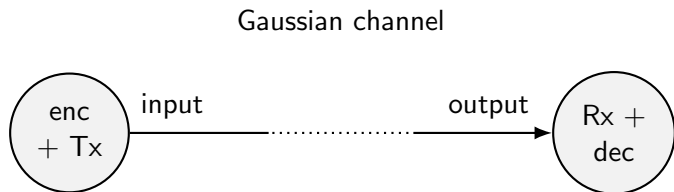
Outline

- 1 Entropy-constrained Gaussian channel
- 2 Moment problems
- 3 Low SNR capacity

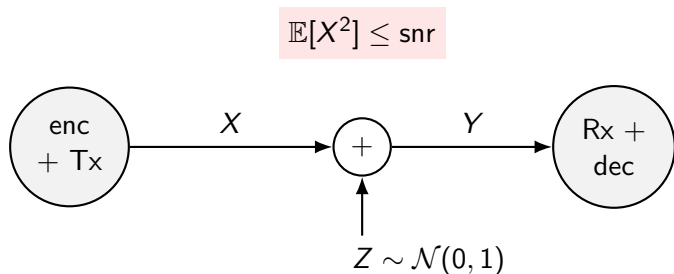
Outline

- 1 Entropy-constrained Gaussian channel
- 2 Moment problems
- 3 Low SNR capacity

Gaussian channel



Gaussian channel

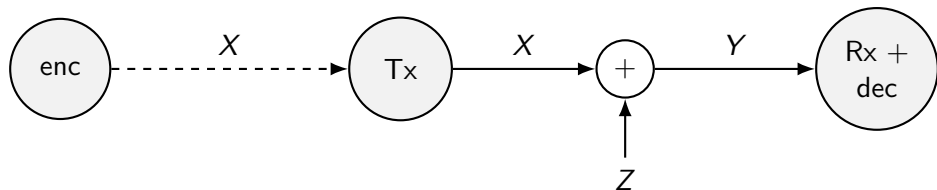


$$C(\text{snr}) = \max_{X: \mathbb{E}[X^2] \leq \text{snr}} I(X; X + Z) = \frac{1}{2} \log(1 + \text{snr})$$

Gaussian channel

finite-capacity noiseless link

$$\mathbb{E}[X^2] \leq \text{snr}$$



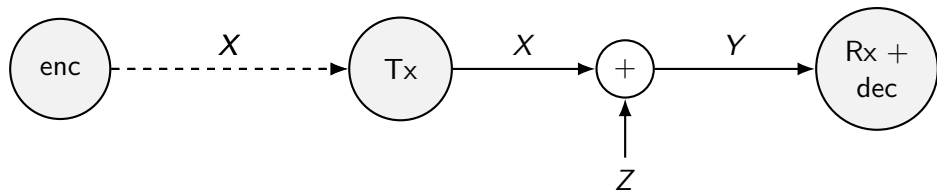
$$C(\text{snr}) = \max_{X: \mathbb{E}[X^2] \leq \text{snr}} I(X; X + Z) = \frac{1}{2} \log(1 + \text{snr})$$

Entropy-constrained Gaussian channel

finite-capacity noiseless link

$$H(X) \leq h$$

$$\mathbb{E}[X^2] \leq \text{snr}$$



$$C(\text{snr}) = \max_{X: \mathbb{E}[X^2] \leq \text{snr}} I(X; X + Z) = \frac{1}{2} \log(1 + \text{snr})$$

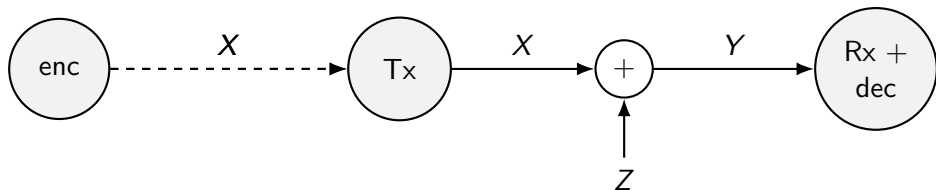
$$C_H(h, \text{snr}) = \max_{\substack{X: \mathbb{E}[X^2] \leq \text{snr} \\ H(X) \leq h}} I(X; X + Z)$$

Entropy-constrained Gaussian channel

finite-capacity noiseless link

$$H(X) \leq h$$

$$\mathbb{E}[X^2] \leq \text{snr}$$



$$C(\text{snr}) = \max_{X: \mathbb{E}[X^2] \leq \text{snr}} I(X; X + Z) = \frac{1}{2} \log(1 + \text{snr})$$

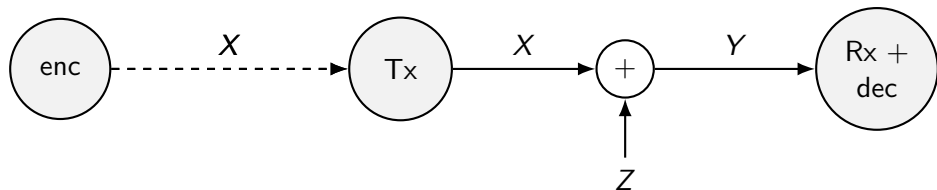
$$C_H(h, \text{snr}) = \max_{\substack{X: \mathbb{E}[X^2] \leq \text{snr} \\ H(X) \leq h}} I(X; X + Z) = \max_{\substack{X: \mathbb{E}[X^2] \leq 1 \\ H(X) \leq h}} I(X; \sqrt{\text{snr}} X + Z)$$

Entropy-constrained Gaussian channel

finite-capacity noiseless link

$$H(X) \leq h$$

$$\mathbb{E}[X^2] \leq \text{snr}$$



$$C(\text{snr}) = \max_{X: \mathbb{E}[X^2] \leq \text{snr}} I(X; X + Z) = \frac{1}{2} \log(1 + \text{snr})$$

$$C_H(h, \text{snr}) = \max_{\substack{X: \mathbb{E}[X^2] \leq \text{snr} \\ H(X) \leq h}} I(X; X + Z) = \max_{\substack{X: \mathbb{E}[X^2] \leq 1 \\ H(X) \leq h}} \underbrace{I(X; \sqrt{\text{snr}}X + Z)}_{I(X, \text{snr})}$$

Approximation perspective

Approximation perspective

- wlog let $\mathbb{E}[X] = 0$, $\mathbb{E}[X^2] = 1$; $G \sim \mathcal{N}(0, 1)$ independent of Z

Approximation perspective

- wlog let $\mathbb{E}[X] = 0$, $\mathbb{E}[X^2] = 1$; $G \sim \mathcal{N}(0, 1)$ independent of Z
- $I(G, \text{snr}) - I(X, \text{snr}) = D(\sqrt{\text{snr}}X + Z \parallel \sqrt{\text{snr}}G + Z)$

Approximation perspective

- wlog let $\mathbb{E}[X] = 0$, $\mathbb{E}[X^2] = 1$; $G \sim \mathcal{N}(0, 1)$ independent of Z
- $I(G, \text{snr}) - I(X, \text{snr}) = D(\sqrt{\text{snr}}X + Z \parallel \sqrt{\text{snr}}G + Z)$
- $C(\text{snr}) - C_H(h, \text{snr}) = \min_{X: \substack{\mathbb{E}[X^2] \leq 1, \\ H(X) \leq h}} D(\sqrt{\text{snr}}X + Z \parallel \sqrt{\text{snr}}G + Z)$

Approximation perspective

- wlog let $\mathbb{E}[X] = 0$, $\mathbb{E}[X^2] = 1$; $G \sim \mathcal{N}(0, 1)$ independent of Z
- $I(G, \text{snr}) - I(X, \text{snr}) = D(\sqrt{\text{snr}}X + Z \parallel \sqrt{\text{snr}}G + Z)$
- $C(\text{snr}) - C_H(h, \text{snr}) = \min_{X: \substack{\mathbb{E}[X^2] \leq 1, \\ H(X) \leq h}} D(\sqrt{\text{snr}}X + Z \parallel \sqrt{\text{snr}}G + Z)$
- optimal distribution at h : discrete X with $H(X) \leq h$ that is closest to $\mathcal{N}(0, 1 + \text{snr})$ after “Gaussian smoothing”

Estimation perspective

Estimation perspective

- MMSE of estimating X from $Y = \sqrt{\text{snr}}X + Z$:

$$\text{mmse}(X, \text{snr}) = \mathbb{E} [(X - \mathbb{E}[X | Y])^2]$$

Estimation perspective

- MMSE of estimating X from $Y = \sqrt{\text{snr}}X + Z$:

$$\text{mmse}(X, \text{snr}) = \mathbb{E} [(X - \mathbb{E}[X | Y])^2]$$

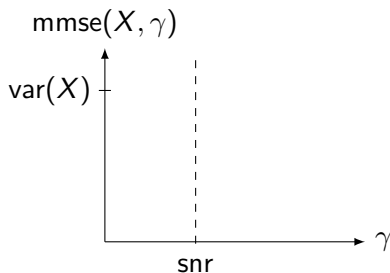
- I-MMSE relationship: $I(X, \text{snr}) = \frac{1}{2} \int_0^{\text{snr}} \text{mmse}(X, \gamma) d\gamma, \quad H(X) = I(X, \infty)$

Estimation perspective

- MMSE of estimating X from $Y = \sqrt{\text{snr}}X + Z$:

$$\text{mmse}(X, \text{snr}) = \mathbb{E} [(X - \mathbb{E}[X | Y])^2]$$

- I-MMSE relationship: $I(X, \text{snr}) = \frac{1}{2} \int_0^{\text{snr}} \text{mmse}(X, \gamma) d\gamma$, $H(X) = I(X, \infty)$

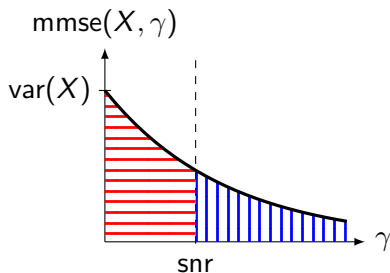


Estimation perspective

- MMSE of estimating X from $Y = \sqrt{\text{snr}}X + Z$:

$$\text{mmse}(X, \text{snr}) = \mathbb{E} [(X - \mathbb{E}[X | Y])^2]$$

- I-MMSE relationship: $I(X, \text{snr}) = \frac{1}{2} \int_0^{\text{snr}} \text{mmse}(X, \gamma) d\gamma$, $H(X) = I(X, \infty)$

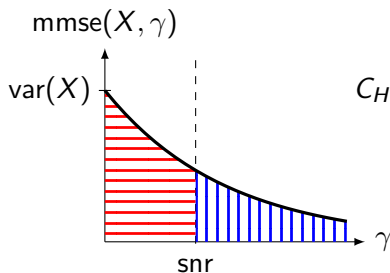


Estimation perspective

- MMSE of estimating X from $Y = \sqrt{\text{snr}}X + Z$:

$$\text{mmse}(X, \text{snr}) = \mathbb{E} [(X - \mathbb{E}[X | Y])^2]$$

- I-MMSE relationship: $I(X, \text{snr}) = \frac{1}{2} \int_0^{\text{snr}} \text{mmse}(X, \gamma) d\gamma$, $H(X) = I(X, \infty)$



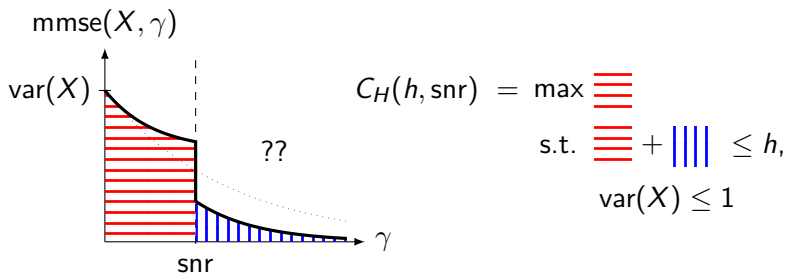
$$C_H(h, \text{snr}) = \max \text{ (red horizontal lines) } \\ \text{s.t. } \text{ (red horizontal lines) } + \text{ (blue vertical lines) } \leq h, \\ \text{var}(X) \leq 1$$

Estimation perspective

- MMSE of estimating X from $Y = \sqrt{\text{snr}}X + Z$:

$$\text{mmse}(X, \text{snr}) = \mathbb{E} [(X - \mathbb{E}[X | Y])^2]$$

- I-MMSE relationship: $I(X, \text{snr}) = \frac{1}{2} \int_0^{\text{snr}} \text{mmse}(X, \gamma) d\gamma$, $H(X) = I(X, \infty)$

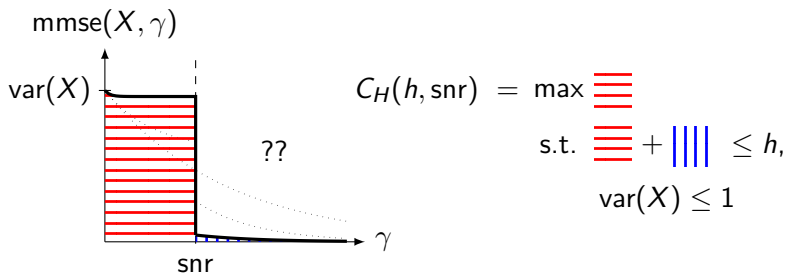


Estimation perspective

- MMSE of estimating X from $Y = \sqrt{\text{snr}}X + Z$:

$$\text{mmse}(X, \text{snr}) = \mathbb{E} [(X - \mathbb{E}[X | Y])^2]$$

- I-MMSE relationship: $I(X, \text{snr}) = \frac{1}{2} \int_0^{\text{snr}} \text{mmse}(X, \gamma) d\gamma$, $H(X) = I(X, \infty)$

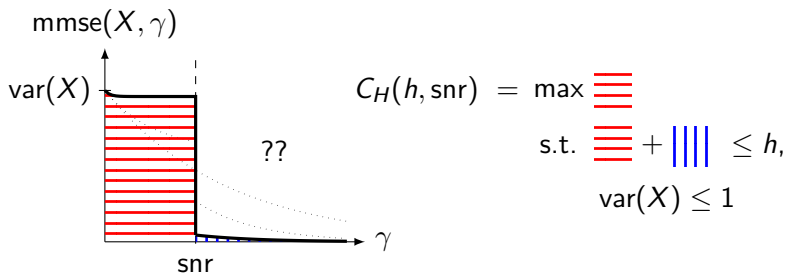


Estimation perspective

- MMSE of estimating X from $Y = \sqrt{\text{snr}}X + Z$:

$$\text{mmse}(X, \text{snr}) = \mathbb{E} [(X - \mathbb{E}[X | Y])^2]$$

- I-MMSE relationship: $I(X, \text{snr}) = \frac{1}{2} \int_0^{\text{snr}} \text{mmse}(X, \gamma) d\gamma$, $H(X) = I(X, \infty)$



- optimal distribution at snr : indistinguishable at $\text{SNR} < \text{snr}$, distinguishable at $\text{SNR} > \text{snr}$

Low SNR asymptotics via Taylor expansion

Low SNR asymptotics via Taylor expansion

- $\mathbb{E}[X^{2n}] < \infty$: $I(X, \text{snr}) = \sum_{i=1}^{n-1} a_{i,X} \text{snr}^i + r_{n,X} \text{snr}^n$

Low SNR asymptotics via Taylor expansion

- $\mathbb{E}[X^{2n}] < \infty$: $I(X, \text{snr}) = \sum_{i=1}^{n-1} a_{i,X} \text{snr}^i + r_{n,X} \text{snr}^n$
- $a_{k,X}$ is a polynomial of $\{\mathbb{E}[X^n] : n = 1, \dots, k\}$ for $k \geq 2$

Low SNR asymptotics via Taylor expansion

- $\mathbb{E}[X^{2n}] < \infty$: $I(X, \text{snr}) = \sum_{i=1}^{n-1} a_{i,X} \text{snr}^i + r_{n,X} \text{snr}^n$
- $a_{k,X}$ is a polynomial of $\{\mathbb{E}[X^n] : n = 1, \dots, k\}$ for $k \geq 2$
- for X with $\mathbb{E}[X^{2(k+2)}] < \infty$:

Low SNR asymptotics via Taylor expansion

- $\mathbb{E}[X^{2n}] < \infty$: $I(X, \text{snr}) = \sum_{i=1}^{n-1} a_{i,X} \text{snr}^i + r_{n,X} \text{snr}^n$
- $a_{k,X}$ is a polynomial of $\{\mathbb{E}[X^n] : n = 1, \dots, k\}$ for $k \geq 2$
- for X with $\mathbb{E}[X^{2(k+2)}] < \infty$: if $\mathbb{E}[X^n] = \mathbb{E}[G^n]$, $n = 1, \dots, k$ and $\mathbb{E}[X^{k+1}] \neq \mathbb{E}[G^{k+1}]$,

Low SNR asymptotics via Taylor expansion

- $\mathbb{E}[X^{2n}] < \infty$: $I(X, \text{snr}) = \sum_{i=1}^{n-1} a_{i,X} \text{snr}^i + r_{n,X} \text{snr}^n$
- $a_{k,X}$ is a polynomial of $\{\mathbb{E}[X^n] : n = 1, \dots, k\}$ for $k \geq 2$
- for X with $\mathbb{E}[X^{2(k+2)}] < \infty$: if $\mathbb{E}[X^n] = \mathbb{E}[G^n]$, $n = 1, \dots, k$ and $\mathbb{E}[X^{k+1}] \neq \mathbb{E}[G^{k+1}]$, then $I(G, \text{snr}) - I(X, \text{snr}) = \Theta(\text{snr}^{k+1})$ as $\text{snr} \rightarrow 0$

Low SNR asymptotics via Taylor expansion

- $\mathbb{E}[X^{2n}] < \infty$: $I(X, \text{snr}) = \sum_{i=1}^{n-1} a_{i,X} \text{snr}^i + r_{n,X} \text{snr}^n$
- $a_{k,X}$ is a polynomial of $\{\mathbb{E}[X^n] : n = 1, \dots, k\}$ for $k \geq 2$
- for X with $\mathbb{E}[X^{2(k+2)}] < \infty$: if $\mathbb{E}[X^n] = \mathbb{E}[G^n]$, $n = 1, \dots, k$ and $\mathbb{E}[X^{k+1}] \neq \mathbb{E}[G^{k+1}]$, then $I(G, \text{snr}) - I(X, \text{snr}) = \Theta(\text{snr}^{k+1})$ as $\text{snr} \rightarrow 0$
- $k_h :=$ maximum number of moments of G matched by X with $H(X) \leq h$

Low SNR asymptotics via Taylor expansion

- $\mathbb{E}[X^{2n}] < \infty$: $I(X, \text{snr}) = \sum_{i=1}^{n-1} a_{i,X} \text{snr}^i + r_{n,X} \text{snr}^n$
- $a_{k,X}$ is a polynomial of $\{\mathbb{E}[X^n] : n = 1, \dots, k\}$ for $k \geq 2$
- for X with $\mathbb{E}[X^{2(k+2)}] < \infty$: if $\mathbb{E}[X^n] = \mathbb{E}[G^n]$, $n = 1, \dots, k$ and $\mathbb{E}[X^{k+1}] \neq \mathbb{E}[G^{k+1}]$, then $I(G, \text{snr}) - I(X, \text{snr}) = \Theta(\text{snr}^{k+1})$ as $\text{snr} \rightarrow 0$
- $k_h :=$ maximum number of moments of G matched by X with $H(X) \leq h$
- $C(\text{snr}) - C_H(h, \text{snr}) = \mathcal{O}(\text{snr}^{k_h+1})$

Low SNR asymptotics via Taylor expansion

- $\mathbb{E}[X^{2n}] < \infty$: $I(X, \text{snr}) = \sum_{i=1}^{n-1} a_{i,X} \text{snr}^i + r_{n,X} \text{snr}^n$
- $a_{k,X}$ is a polynomial of $\{\mathbb{E}[X^n] : n = 1, \dots, k\}$ for $k \geq 2$
- for X with $\mathbb{E}[X^{2(k+2)}] < \infty$: if $\mathbb{E}[X^n] = \mathbb{E}[G^n]$, $n = 1, \dots, k$ and $\mathbb{E}[X^{k+1}] \neq \mathbb{E}[G^{k+1}]$, then $I(G, \text{snr}) - I(X, \text{snr}) = \Theta(\text{snr}^{k+1})$ as $\text{snr} \rightarrow 0$
- $k_h :=$ maximum number of moments of G matched by X with $H(X) \leq h$
- $C(\text{snr}) - C_H(h, \text{snr}) = \mathcal{O}(\text{snr}^{k_h+1})$
(\mathcal{O} instead of Θ to allow for X with $\mathbb{E}[X^{2(k_h+2)}] = \infty$)

Outline

- 1 Entropy-constrained Gaussian channel
- 2 **Moment problems**
- 3 Low SNR capacity

Classical moment problem

Classical moment problem

- Q: Given s_1, s_2, s_3, \dots , does there exist X on \mathbb{R} such that $\mathbb{E}[X^n] = s_n$ for $n = 1, 2, 3, \dots$?

Classical moment problem

- Q: Given s_1, s_2, s_3, \dots , does there exist X on \mathbb{R} such that $\mathbb{E}[X^n] = s_n$ for $n = 1, 2, 3, \dots$?

A: iff $H_n(s_1, \dots, s_{2n}) = \begin{pmatrix} 1 & s_1 & \dots & s_n \\ s_1 & s_2 & \dots & s_{n+1} \\ \vdots & \vdots & \ddots & \vdots \\ s_n & s_{n+1} & \dots & s_{2n} \end{pmatrix} \succeq 0$ for $n = 1, 2, 3, \dots$

Classical moment problem

- Q: Given s_1, s_2, s_3, \dots , does there exist X on \mathbb{R} such that $\mathbb{E}[X^n] = s_n$ for $n = 1, 2, 3, \dots$?

A: iff $H_n(s_1, \dots, s_{2n}) = \begin{pmatrix} 1 & s_1 & \dots & s_n \\ s_1 & s_2 & \dots & s_{n+1} \\ \vdots & \vdots & \ddots & \vdots \\ s_n & s_{n+1} & \dots & s_{2n} \end{pmatrix} \succeq 0$ for $n = 1, 2, 3, \dots$;

infinite support iff $H_n \succ 0$ for all n

Truncated moment problem

- Q: Given s_1, s_2, s_3, \dots , does there exist X on \mathbb{R} such that $\mathbb{E}[X^n] = s_n$ for $n = 1, 2, 3, \dots$?

A: iff $H_n(s_1, \dots, s_{2n}) = \begin{pmatrix} 1 & s_1 & \dots & s_n \\ s_1 & s_2 & \dots & s_{n+1} \\ \vdots & \vdots & \ddots & \vdots \\ s_n & s_{n+1} & \dots & s_{2n} \end{pmatrix} \succeq 0$ for $n = 1, 2, 3, \dots$;

infinite support iff $H_n \succ 0$ for all n

- Q: Given $s_1, s_2, s_3, \dots, s_k$, does there exist X on \mathbb{R} such that $\mathbb{E}[X^n] = s_n$ for $n = 1, 2, 3, \dots, k$?

Truncated moment problem

- Q: Given s_1, s_2, s_3, \dots , does there exist X on \mathbb{R} such that $\mathbb{E}[X^n] = s_n$ for $n = 1, 2, 3, \dots$?

A: iff $H_n(s_1, \dots, s_{2n}) = \begin{pmatrix} 1 & s_1 & \dots & s_n \\ s_1 & s_2 & \dots & s_{n+1} \\ \vdots & \vdots & \ddots & \vdots \\ s_n & s_{n+1} & \dots & s_{2n} \end{pmatrix} \succeq 0$ for $n = 1, 2, 3, \dots$;

infinite support iff $H_n \succ 0$ for all n

- Q: Given $s_1, s_2, s_3, \dots, s_k$, does there exist X on \mathbb{R} such that $\mathbb{E}[X^n] = s_n$ for $n = 1, 2, 3, \dots, k$?

A: iff (k odd)

(k even)

Truncated moment problem

- Q: Given s_1, s_2, s_3, \dots , does there exist X on \mathbb{R} such that $\mathbb{E}[X^n] = s_n$ for $n = 1, 2, 3, \dots$?

A: iff $H_n(s_1, \dots, s_{2n}) = \begin{pmatrix} 1 & s_1 & \dots & s_n \\ s_1 & s_2 & \dots & s_{n+1} \\ \vdots & \vdots & \ddots & \vdots \\ s_n & s_{n+1} & \dots & s_{2n} \end{pmatrix} \succeq 0$ for $n = 1, 2, 3, \dots$;

infinite support iff $H_n \succ 0$ for all n

- Q: Given $s_1, s_2, s_3, \dots, s_k$, does there exist X on \mathbb{R} such that $\mathbb{E}[X^n] = s_n$ for $n = 1, 2, 3, \dots, k$?

A: iff (k odd) there exists \tilde{s}_{k+1} such that $H_{\frac{k+1}{2}}(s_1, \dots, s_k, \tilde{s}_{k+1}) \succeq 0$
(k even)

Truncated moment problem

- Q: Given s_1, s_2, s_3, \dots , does there exist X on \mathbb{R} such that $\mathbb{E}[X^n] = s_n$ for $n = 1, 2, 3, \dots$?

A: iff $H_n(s_1, \dots, s_{2n}) = \begin{pmatrix} 1 & s_1 & \dots & s_n \\ s_1 & s_2 & \dots & s_{n+1} \\ \vdots & \vdots & \ddots & \vdots \\ s_n & s_{n+1} & \dots & s_{2n} \end{pmatrix} \succeq 0$ for $n = 1, 2, 3, \dots$;

infinite support iff $H_n \succ 0$ for all n

- Q: Given $s_1, s_2, s_3, \dots, s_k$, does there exist X on \mathbb{R} such that $\mathbb{E}[X^n] = s_n$ for $n = 1, 2, 3, \dots, k$?

A: iff (k odd) there exists \tilde{s}_{k+1} such that $H_{\frac{k+1}{2}}(s_1, \dots, s_k, \tilde{s}_{k+1}) \succeq 0$

(k even) there exist $\tilde{s}_{k+1}, \tilde{s}_{k+2}$ such that $H_{\frac{k}{2}+1}(s_1, \dots, s_k, \tilde{s}_{k+1}, \tilde{s}_{k+2}) \succeq 0$

Truncated moment problem

- Q: Given s_1, s_2, s_3, \dots , does there exist X on \mathbb{R} such that $\mathbb{E}[X^n] = s_n$ for $n = 1, 2, 3, \dots$?

A: iff $H_n(s_1, \dots, s_{2n}) = \begin{pmatrix} 1 & s_1 & \dots & s_n \\ s_1 & s_2 & \dots & s_{n+1} \\ \vdots & \vdots & \ddots & \vdots \\ s_n & s_{n+1} & \dots & s_{2n} \end{pmatrix} \succeq 0$ for $n = 1, 2, 3, \dots$;

infinite support iff $H_n \succ 0$ for all n

- Q: Given $s_1, s_2, s_3, \dots, s_k$, does there exist X on \mathbb{R} such that $\mathbb{E}[X^n] = s_n$ for $n = 1, 2, 3, \dots, k$?

A: iff (k odd) there exists \tilde{s}_{k+1} such that $H_{\frac{k+1}{2}}(s_1, \dots, s_k, \tilde{s}_{k+1}) \succeq 0$

(k even) there exist $\tilde{s}_{k+1}, \tilde{s}_{k+2}$ such that $H_{\frac{k}{2}+1}(s_1, \dots, s_k, \tilde{s}_{k+1}, \tilde{s}_{k+2}) \succeq 0$;

finite support: at most $\lfloor k/2 \rfloor + 1$ atoms (if it exists)

Outline

- 1 Entropy-constrained Gaussian channel
- 2 Moment problems
- 3 Low SNR capacity

Main result

Theorem

For any continuous W , there exists $\eta_W \in (0, \frac{1}{2})$ such that

Main result

Theorem

For any continuous W , there exists $\eta_W \in (0, \frac{1}{2})$ such that

(i) if X has $\mathbb{E}[X^n] = \mathbb{E}[W^n]$ for $n = 1, 2, 3, 4$, then $H(X) \geq h_2(\eta_W)$.

Main result

Theorem

For any continuous W , there exists $\eta_W \in (0, \frac{1}{2})$ such that

- (i) if X has $\mathbb{E}[X^n] = \mathbb{E}[W^n]$ for $n = 1, 2, 3, 4$, then $H(X) \geq h_2(\eta_W)$.*
- (ii) for any $h > 0$, there is X with $H(X) \leq h$ and $\mathbb{E}[X^n] = \mathbb{E}[W^n]$ for $n = 1, 2, 3$.*

Main result

Theorem

For any continuous W , there exists $\eta_W \in (0, \frac{1}{2})$ such that

- (i) if X has $\mathbb{E}[X^n] = \mathbb{E}[W^n]$ for $n = 1, 2, 3, 4$, then $H(X) \geq h_2(\eta_W)$.
- (ii) for any $h > 0$, there is X with $H(X) \leq h$ and $\mathbb{E}[X^n] = \mathbb{E}[W^n]$ for $n = 1, 2, 3$.

Proof idea



Main result

Theorem

For any continuous W , there exists $\eta_W \in (0, \frac{1}{2})$ such that

- (i) if X has $\mathbb{E}[X^n] = \mathbb{E}[W^n]$ for $n = 1, 2, 3, 4$, then $H(X) \geq h_2(\eta_W)$.
- (ii) for any $h > 0$, there is X with $H(X) \leq h$ and $\mathbb{E}[X^n] = \mathbb{E}[W^n]$ for $n = 1, 2, 3$.

Proof idea

- $H(X) \leq h < \log 2 \iff X = \begin{cases} \tilde{X} & \text{w.p. } \epsilon < 1/2 \\ x_0 & \text{w.p. } 1 - \epsilon > 1/2 \end{cases}$



Main result

Theorem

For any continuous W , there exists $\eta_W \in (0, \frac{1}{2})$ such that

- (i) if X has $\mathbb{E}[X^n] = \mathbb{E}[W^n]$ for $n = 1, 2, 3, 4$, then $H(X) \geq h_2(\eta_W)$.
- (ii) for any $h > 0$, there is X with $H(X) \leq h$ and $\mathbb{E}[X^n] = \mathbb{E}[W^n]$ for $n = 1, 2, 3$.

Proof idea

- $H(X) \leq h < \log 2 \iff X = \begin{cases} \tilde{X} & \text{w.p. } \epsilon < 1/2 \\ x_0 & \text{w.p. } 1 - \epsilon > 1/2 \end{cases}$
- need $\mathbb{E}[\tilde{X}^n] = s_n := \frac{1}{\epsilon}(\mathbb{E}[W^n] - (1 - \epsilon)x_0^n)$



Main result

Theorem

For any continuous W , there exists $\eta_W \in (0, \frac{1}{2})$ such that

- (i) if X has $\mathbb{E}[X^n] = \mathbb{E}[W^n]$ for $n = 1, 2, 3, 4$, then $H(X) \geq h_2(\eta_W)$.
- (ii) for any $h > 0$, there is X with $H(X) \leq h$ and $\mathbb{E}[X^n] = \mathbb{E}[W^n]$ for $n = 1, 2, 3$.

Proof idea

- $H(X) \leq h < \log 2 \iff X = \begin{cases} \tilde{X} & \text{w.p. } \epsilon < 1/2 \\ x_0 & \text{w.p. } 1 - \epsilon > 1/2 \end{cases}$
- need $\mathbb{E}[\tilde{X}^n] = s_n := \frac{1}{\epsilon}(\mathbb{E}[W^n] - (1 - \epsilon)x_0^n)$
- check: s_1, \dots, s_4 “valid” iff $\epsilon > \eta_W$, but s_1, s_2, s_3 always “valid”



Main result

Theorem

For any continuous W , there exists $\eta_W \in (0, \frac{1}{2})$ such that

- (i) if X has $\mathbb{E}[X^n] = \mathbb{E}[W^n]$ for $n = 1, 2, 3, 4$, then $H(X) \geq h_2(\eta_W)$.
- (ii) for any $h > 0$, there is X with $H(X) \leq h$ and $\mathbb{E}[X^n] = \mathbb{E}[W^n]$ for $n = 1, 2, 3$.

Proof idea

- $H(X) \leq h < \log 2 \iff X = \begin{cases} \tilde{X} & \text{w.p. } \epsilon < 1/2 \\ x_0 & \text{w.p. } 1 - \epsilon > 1/2 \end{cases}$
- need $\mathbb{E}[\tilde{X}^n] = s_n := \frac{1}{\epsilon}(\mathbb{E}[W^n] - (1 - \epsilon)x_0^n)$
- check: s_1, \dots, s_4 “valid” iff $\epsilon > \eta_W$, but s_1, s_2, s_3 always “valid”



Corollary

$$\eta_G = \frac{1}{3}$$

Main result

Theorem

For any continuous W , there exists $\eta_W \in (0, \frac{1}{2})$ such that

- (i) if X has $\mathbb{E}[X^n] = \mathbb{E}[W^n]$ for $n = 1, 2, 3, 4$, then $H(X) \geq h_2(\eta_W)$.
- (ii) for any $h > 0$, there is X with $H(X) \leq h$ and $\mathbb{E}[X^n] = \mathbb{E}[W^n]$ for $n = 1, 2, 3$.

Proof idea

- $H(X) \leq h < \log 2 \iff X = \begin{cases} \tilde{X} & \text{w.p. } \epsilon < 1/2 \\ x_0 & \text{w.p. } 1 - \epsilon > 1/2 \end{cases}$
- need $\mathbb{E}[\tilde{X}^n] = s_n := \frac{1}{\epsilon}(\mathbb{E}[W^n] - (1 - \epsilon)x_0^n)$
- check: s_1, \dots, s_4 “valid” iff $\epsilon > \eta_W$, but s_1, s_2, s_3 always “valid”



Corollary

$\eta_G = \frac{1}{3}$, so for $h < h_2(\frac{1}{3})$, as $\text{snr} \rightarrow 0$, $C(\text{snr}) - C_H(h, \text{snr}) = \mathcal{O}(\text{snr}^4)$.

Summary

- entropy-constrained Gaussian channel

Summary

- entropy-constrained Gaussian channel capacity $C_H(h, \text{snr})$

Summary

- entropy-constrained Gaussian channel capacity $C_H(h, \text{snr})$
- $\text{snr} \rightarrow 0$, $0 < h < h_2(\frac{1}{3})$: $C_H(h, \text{snr}) = C(\text{snr}) - \mathcal{O}(\text{snr}^4)$

Summary

- entropy-constrained Gaussian channel capacity $C_H(h, \text{snr})$
- $\text{snr} \rightarrow 0$, $0 < h < h_2(\frac{1}{3})$: $C_H(h, \text{snr}) = C(\text{snr}) - \mathcal{O}(\text{snr}^4)$, via entropy-constrained version of truncated moment problem

Summary

- entropy-constrained Gaussian channel capacity $C_H(h, \text{snr})$
- $\text{snr} \rightarrow 0$, $0 < h < h_2(\frac{1}{3})$: $C_H(h, \text{snr}) = C(\text{snr}) - \mathcal{O}(\text{snr}^4)$, via entropy-constrained version of truncated moment problem
- for any continuous distribution, only three moments can be matched by a discrete distribution of sufficiently small entropy

Summary

- entropy-constrained Gaussian channel capacity $C_H(h, \text{snr})$
- $\text{snr} \rightarrow 0$, $0 < h < h_2(\frac{1}{3})$: $C_H(h, \text{snr}) = C(\text{snr}) - \mathcal{O}(\text{snr}^4)$, via entropy-constrained version of truncated moment problem
- for any continuous distribution, only three moments can be matched by a discrete distribution of sufficiently small entropy
- open:

Summary

- entropy-constrained Gaussian channel capacity $C_H(h, \text{snr})$
- $\text{snr} \rightarrow 0$, $0 < h < h_2(\frac{1}{3})$: $C_H(h, \text{snr}) = C(\text{snr}) - \mathcal{O}(\text{snr}^4)$, via entropy-constrained version of truncated moment problem
- for any continuous distribution, only three moments can be matched by a discrete distribution of sufficiently small entropy
- open:
 - structure of capacity-achieving distributions

Summary

- entropy-constrained Gaussian channel capacity $C_H(h, \text{snr})$
- $\text{snr} \rightarrow 0$, $0 < h < h_2(\frac{1}{3})$: $C_H(h, \text{snr}) = C(\text{snr}) - \mathcal{O}(\text{snr}^4)$, via entropy-constrained version of truncated moment problem
- for any continuous distribution, only three moments can be matched by a discrete distribution of sufficiently small entropy
- open:
 - structure of capacity-achieving distributions
 - $C_H(h, \text{snr})$ for other h, snr

Summary

- entropy-constrained Gaussian channel capacity $C_H(h, \text{snr})$
- $\text{snr} \rightarrow 0$, $0 < h < h_2(\frac{1}{3})$: $C_H(h, \text{snr}) = C(\text{snr}) - \mathcal{O}(\text{snr}^4)$, via entropy-constrained version of truncated moment problem
- for any continuous distribution, only three moments can be matched by a discrete distribution of sufficiently small entropy
- open:
 - structure of capacity-achieving distributions
 - $C_H(h, \text{snr})$ for other h, snr
 - insights from/to estimation problems?

Summary

- entropy-constrained Gaussian channel capacity $C_H(h, \text{snr})$
- $\text{snr} \rightarrow 0$, $0 < h < h_2(\frac{1}{3})$: $C_H(h, \text{snr}) = C(\text{snr}) - \mathcal{O}(\text{snr}^4)$, via entropy-constrained version of truncated moment problem
- for any continuous distribution, only three moments can be matched by a discrete distribution of sufficiently small entropy
- open:
 - structure of capacity-achieving distributions
 - $C_H(h, \text{snr})$ for other h, snr
 - insights from/to estimation problems?

Thank you!