# Multi-Rate Control over AWGN Channels via Analog Coding

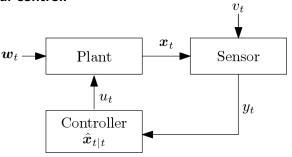
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Joint work with: Gustav M. Pettersson, KTH Babak Hassibi and Victoria Kostina, Caltech

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## Networked Control vs. Traditional Control

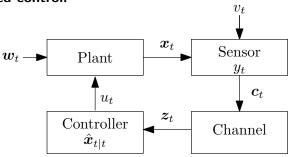
#### Traditional control:



- Observer and controller are co-located.
- Classical systems are hardwired and well crafted

#### Networked Control vs. Traditional Control

#### **Networked control:**



- Observer and controller are not co-located: connected through noisy link
- Suitable for new remote applications (e.g., remote surgery, self-driving cars)



#### Linear quadratic Gaussian (LQG) system

$$egin{aligned} oldsymbol{x}_{t+1} &= oldsymbol{\mathsf{A}} oldsymbol{x}_t + oldsymbol{\mathsf{B}} oldsymbol{u}_t + oldsymbol{\mathsf{W}}_t, & oldsymbol{w}_t \sim \text{ i.i.d. } \mathcal{N}(0, oldsymbol{\mathsf{W}}) \ oldsymbol{y}_t &= oldsymbol{\mathsf{C}} oldsymbol{x}_t + oldsymbol{\mathsf{v}}_t, & oldsymbol{v}_t \sim \text{ i.i.d. } \mathcal{N}(0, oldsymbol{\mathsf{W}}) \end{aligned}$$

#### Additive white Gaussian noise (AWGN) channel

$$oldsymbol{b}_t = oldsymbol{\mathsf{H}} oldsymbol{a}_t + oldsymbol{n}_t, \qquad oldsymbol{n}_t \sim \text{ i.i.d. } \mathcal{N}\left(0, oldsymbol{\mathsf{N}}
ight)$$

• Power constraint:  $\mathbb{E}\left[\boldsymbol{a}_{t}^{T}\boldsymbol{a}_{t}\right] = \sum_{i} \mathbb{E}\left[a_{t;i}^{2}\right] \leq P \cdot \mathsf{length}(\boldsymbol{a}_{t})$ 

#### LQG cost

$$J = \sum_{t=1}^{T} \left[ \boldsymbol{x}_{t}^{T} \mathbf{Q} \boldsymbol{x}_{t} + \boldsymbol{u}_{t}^{T} \mathbf{R} \boldsymbol{u}_{t} \right] + \boldsymbol{x}_{T+1}^{T} \mathbf{F} \boldsymbol{x}_{T+1}$$



#### Scalar Linear quadratic Gaussian (LQG) system

$$x_{t+1} = x_t + u_t + w_t,$$
  $w_t \sim \text{ i.i.d. } \mathcal{N}(0, W)$   
 $y_t = x_t + v_t,$   $v_t \sim \text{ i.i.d. } \mathcal{N}(0, V)$ 

#### Scalar Additive white Gaussian noise (AWGN) channel

$$b_t = a_t + n_t,$$
  $n_t \sim \text{ i.i.d. } \mathcal{N}(0, N)$ 

• Power constraint:  $\mathbb{E}\left[a_t^2\right] \leq P$ ;

#### LQG cost

$$J = \sum_{t=1}^{T} \left[ Qx_t^2 + Ru_t^2 \right] + Fx_{T+1}^2$$



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#### Scalar Additive white Gaussian noise (AWGN) channel

$$b_t = a_t + n_t,$$
  $n_t \sim \text{ i.i.d. } \mathcal{N}(0, N)$ 

• Power constraint:  $\mathbb{E}\left[a_t^2\right] \leq P$ ; w.l.o.g. P=1,  $N=1/\mathsf{SNR}$ 

#### LQG cost

$$J = \sum_{t=1}^{T} \left[ Qx_t^2 + Ru_t^2 \right] + Fx_{T+1}^2$$

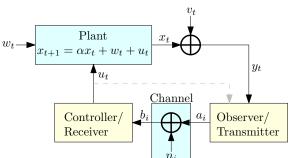


#### Scalar LQG system

$$x_{t+1} = x_t + u_t + w_t$$
$$y_t = x_t + v_t$$

#### Scalar AWGN channel

$$b_t = a_t \left( y^t, u^{t-1} \right) + n_t$$
• Power constraint:  $\mathbb{E} \left[ a_t^2 \right] \leq P$ 



#### **Control sampling rate** $\neq$ **Communication signaling rate!**

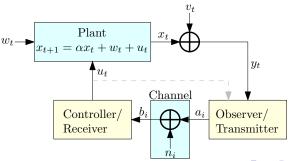
#### Scalar LQG system

$$x_{t+1} = x_t + u_t + w_t$$
$$y_t = x_t + v_t$$

#### Scalar AWGN channel

$$b_i = a_i + n_i$$

• Power constraint:  $\mathbb{E}\left[a_i^2\right] \leq P$ 



## Control Sampling Rate vs. Communication Signaling Rate

- How fast the plant dynamic is ⇒ Control sampling rate
- Bandwidth available ⇒ Communication signaling rate
- Communication rate can be much higher in practice

How to benefit from excess signaling rate (bandwidth)?

## Information-theoretic separation

- Requires large blocks (delay!) of source samples and channel uses
- Suboptimal for control!
- Requires codes with strong "anytime reliability" properties
   [Schulman IT'96][Sahai-Mitter IT'06][Sukhavasi-Hassibi AC'16]
- Problematic in practice: Convolutional code with infinite memory [Kh.-Halbawi-Hassibi ISIT'16]

#### **Packeting**

- Assumes communication rate ≫ control rate, very good SNR
- Problem reduces to control-oriented quantization [My second talk today]
- Bad channel events are translated to packet drops / delays



## Networked Control Approaches: Joint Source-Channel Coding (JSCC)

- What to do when control and communication rates are close?
- Can we do better than IT-separation?

#### Less familiar IT avenue

- Low-delay joint source—channel coding (JSCC)
- Control sample corresponds to source sample

One AWGN channel use per one control sample

#### 1: 1 Optimal JSCC [Goblick IT'65]

- 1:1 optimal JSCC distortion = n:n optimal JSCC distortion
  - No loss of performance
  - Analog scheme is optimal:  $a_t = \sqrt{\frac{P}{P_x}} x_t$

#### Scheme

 For simplicity, assume fully-observable case (can be extended to partially-observable case)

#### Observer/Transmitter:

- ullet Generates the "source" signal:  $s_t = x_t \hat{x}_{t|t-1} = ilde{x}_{t|t-1}$
- Adjusts power and transmits:  $a_t = s_t / \sqrt{P_{t|t-1}}$

#### Controller/Receiver:

- Receives  $b_t = a_t + n_t = \tilde{x}_{t|t-1} / \sqrt{P_{t|t-1}} + n_t$
- $\bullet \text{ Applies Kalman filtering: } \begin{cases} \hat{x}_{t|t} = \hat{x}_{t|t-1} + \sqrt{P_{t|t-1}} \frac{\mathsf{SNR}}{1+\mathsf{SNR}} b_t \\ \hat{x}_{t|t-1} = \alpha \hat{x}_{t-1|t-1} + u_{t-1} \end{cases}$
- Generates LQG control signal:  $u_t = -L_t \hat{x}_{t|t}$



#### Motivation Model NCS JSCC approach Discussion 1: 1 JSCC: Rate-Matched Case

We reduced problem to that of classical LQG control

#### LQR coefficients

$$L_t = \frac{\alpha S_{t+1}}{S_{t+1} + R},$$

$$S_t = \frac{\alpha^2 R S_{t+1}}{S_{t+1} + R} + Q,$$

$$S_T = F.$$

#### Partially-observable case

Scheme can be extended to partially-observable case

• Generates state estimators at the transmitter  $\hat{x}_{t|t}^t$ (in addition to  $\hat{x}_{t|t}^r$  at the receiver)



#### LQG cost

- This schemes achieves optimal LQG cost
- Formally proved by applying
  - Shannon's lower bound
  - Entropy-power inequality
  - Tightness of both in Gaussian case
  - Optimality of "1 : 1 JSCC" scheme in the Gaussian case in the dynamic-programming solution (extension of [Kostina-Hassibi Allerton'16])
- Recovers results of [Freudenberg-Middleton-Solo AC'10] as a special case

#### Conclusion: No coding is needed!



#### Infinite-horizon steady-state average-stage LQG cost

$$ar{J}^{\mathrm{r}} \leq ar{J}^{\mathrm{t}} + rac{Q + \left(lpha^2 - 1
ight)S}{1 + \mathsf{SNR} - lpha^2}W$$
 $ar{J}^{\mathrm{t}} = SW$ 

• S is the positive solution of the DARE

$$S^{2} - [Q + (\alpha^{2} - 1) R] S - QR = 0$$

- System is stabilizabile if and only if SNR  $> \alpha^2 1$
- This is in stark contrast to classical LQG

Conclusion: No coding is needed!



#### Infinite-horizon LQG cost: Partially-observable case

$$ar{J^{t}} \leq ar{J^{t}} + rac{Q + \left(lpha^{2} - 1
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 $ar{J^{t}} = S\left(P_{t}^{t} - ar{P}_{t}^{t}
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- System is stabilizabile if and only if SNR  $> \alpha^2 1$
- $\bullet \ P^t \triangleq \lim_{t \to \infty} P^t_{t+1|t} \ , \qquad \bar{P}^t \triangleq \lim_{t \to \infty} P^t_{t|t}$
- $P^t$  is the positive solution of the DARE  $\left(P^t\right)^2 \left[\left(\alpha^2 1\right)V + W\right]P^t VW = 0, \ \bar{P}^t = \frac{P^tV}{P^t + V}$

#### Conclusion: No coding is needed!



#### Infinite-horizon LQG cost: Partially-observable case

$$\begin{split} \bar{J}^{t} &\leq \bar{J}^{t} + \frac{Q + \left(\alpha^{2} - 1\right)S}{1 + \mathsf{SNR} - \alpha^{2}} (P_{t}^{t} - \bar{P}_{t}^{t}) \\ \bar{J}^{t} &= S(P_{t}^{t} - \bar{P}_{t}^{t}) + Q\bar{P}_{t}^{t} \end{split}$$

• S is the positive solution of the DARE

$$S^{2} - [Q + (\alpha^{2} - 1) R] S - QR = 0$$

- System is stabilizabile if and only if SNR  $> \alpha^2 1$
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#### What about 1 : 2 case?



Two AWGN channel uses per one control sample

#### Naïve scheme: Repetition

Observer/Transmitter: 
$$a_{t;1} = a_{t;2} = \tilde{x}_t / \sqrt{P_{t|t-1}}$$

Controller/Receiver: 
$$b_t^{\text{eff}} = \frac{b_{t;1} + b_{t;2}}{2}$$

- Reduces to 1:1 JSCC with  $SNR^{eff} = 2SNR$
- 3dB improvement comes from doubling total transmit power
- Same improvement is attained by
  - Using 2P during first channel use
  - Remaining silent during second channel use
- No real improvement due to extra degree of freedom...

#### Can we do better?



## Infinite blocklength: "n : 2n JSCC" for $n \to \infty$ [Shannon '48]

$$1 + \mathsf{SNR}^{\mathrm{eff}} = (1 + \mathsf{SNR})^2$$

• Much better than  $SNR_{naive}^{eff} = 2SNR$  at high SNR

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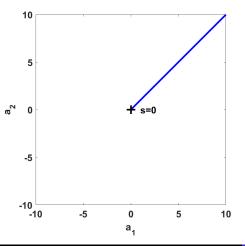
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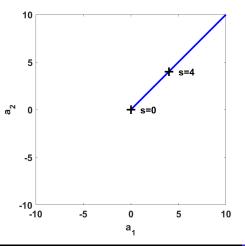
#### What about 1: 2 JSCC?

Non-linear mappings can do better! [Kotel'nikov '47][Shannon '49]

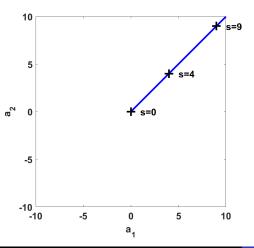
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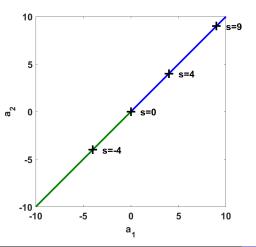
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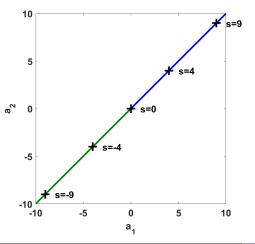
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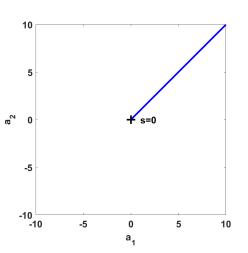


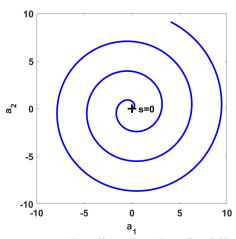
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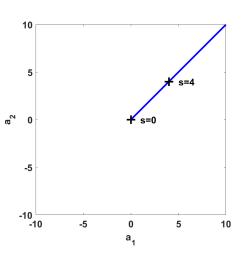
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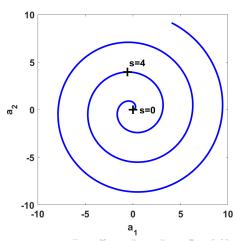




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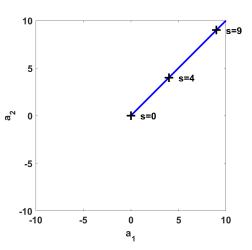
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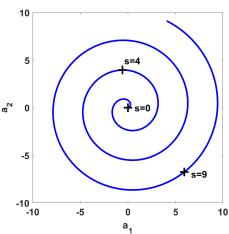




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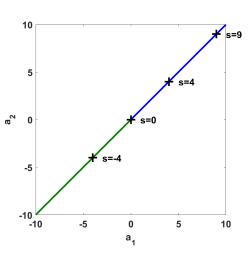
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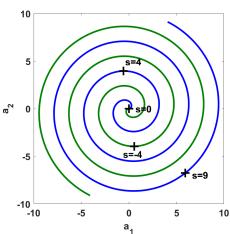




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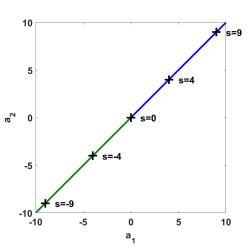
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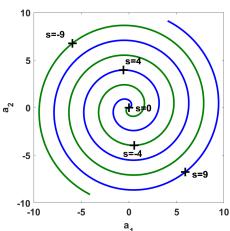




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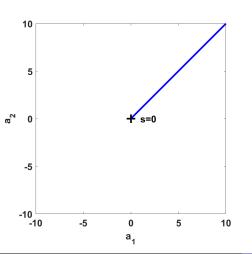
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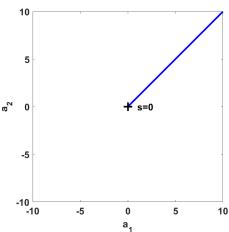




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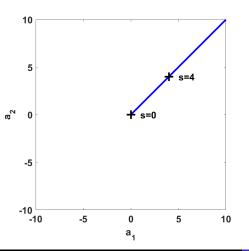
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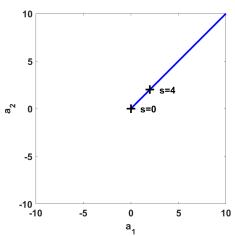




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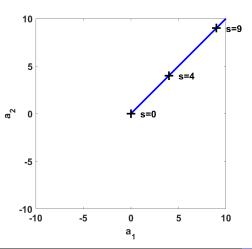
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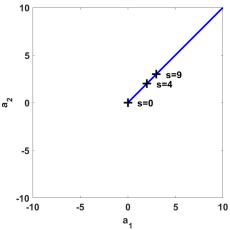




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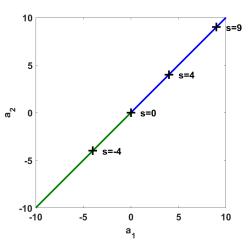
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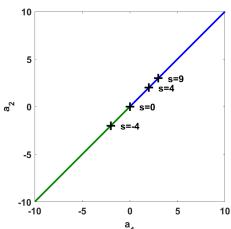




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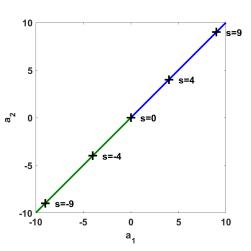
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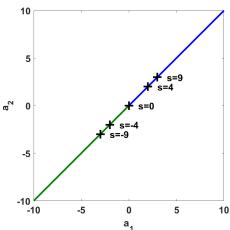




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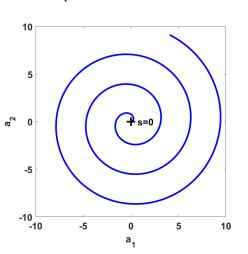
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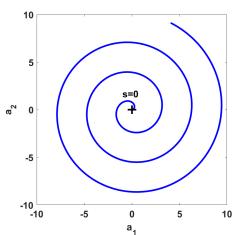




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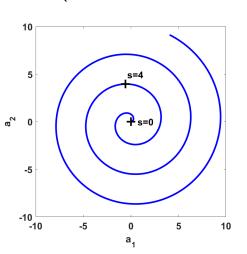
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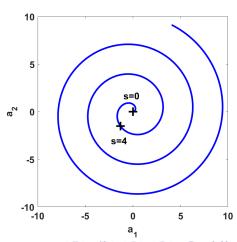




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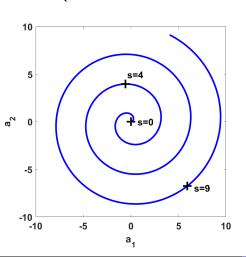
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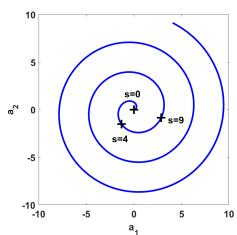




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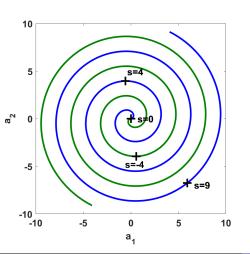
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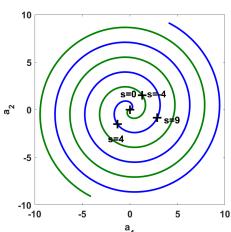




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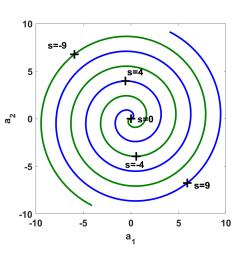
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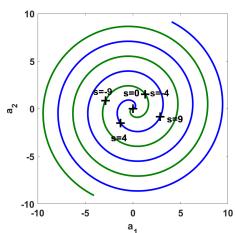


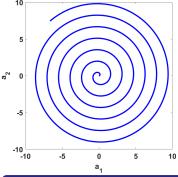


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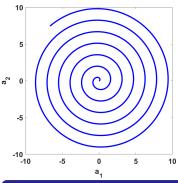


- Small distance between branches
  - ⇒ better for "weak noise"
- Large distance between branches
  - ⇒ better for "strong noise"

### Standard spiral

$$\begin{cases} a_1(s) \propto s \cos(\omega s) &= |s| \cos(\omega |s|) \operatorname{sign}(s) \\ a_2(s) \propto s \sin(\omega s) \operatorname{sign}(s) &= |s| \sin(\omega |s|) \operatorname{sign}(s) \end{cases}$$





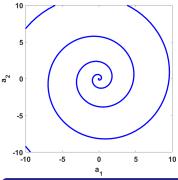
- Small distance between branches
  - ⇒ better for "weak noise"
- Large distance between branches
  - ⇒ better for "strong noise"

#### Stretched-source spiral

Stretch input before mapping to spiral:  $s \to |s|^{\lambda} \operatorname{sign}(s)$ 

$$\begin{cases} a_1(s) \propto |s|^{\lambda} \cos \left(\omega |s|^{\lambda}\right) \operatorname{sign}(s) \\ a_2(s) \propto |s|^{\lambda} \sin \left(\omega |s|^{\lambda}\right) \operatorname{sign}(s) \end{cases}$$





#### Control requirements

- Small distance between branches
  - $\Rightarrow$  better for "weak noise"
- Large distance between branches
  - ⇒ better for "strong noise"

### Bounded average distortion given any input

Avoid increase in distortion with  $|s| \Rightarrow$  Slower rotation with |s|

$$\begin{cases} a_1(s) \propto |s|^{\lambda\beta} \cos \left(\omega |s|^{\lambda}\right) \operatorname{sign}(s) \\ a_2(s) \propto |s|^{\lambda\beta} \sin \left(\omega |s|^{\lambda}\right) \operatorname{sign}(s) \end{cases}$$

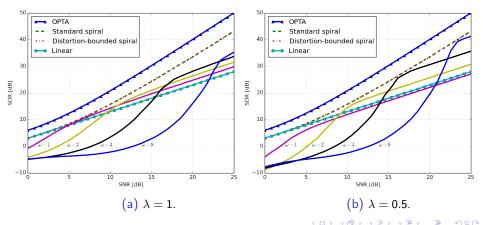




#### **Control requirements**

- Small distance between branches
  - ⇒ better for "weak noise"
- Large distance between branches
  - ⇒ better for "strong noise"

- Average distortion given (almost) any s needs to be small!
- E.g., transmitters that truncate the signal do not perform well (avalanche effect)



#### Inner bound: Black-box approach

Assume a JSCC scheme with bounded distortion  $D = \frac{1}{\mathsf{SNR}^{\mathsf{eff}}}, \forall s$ . Then,

$$ar{J^{ ext{r}}} \leq ar{J^{ ext{t}}} + rac{Q + \left(lpha^2 - 1
ight)S}{1 + \mathsf{SNR}^{ ext{eff}} - lpha^2} (P_t^t - ar{P}_t^t)$$

## Outer bound: Extension of [Kostina-Hassibi Allerton'16]

$$ar{J}^{t} \leq ar{J}^{t} + rac{Q + \left(lpha^{2} - 1
ight)S}{1 + \mathsf{SNR}_{p 
ightarrow \infty}^{\mathrm{eff}} - lpha^{2}} (P_{t}^{t} - ar{P}_{t}^{t})$$

- $1 + \mathsf{SNR}_{n \to \infty}^{\text{eff}} = (1 + \mathsf{SNR})^2$
- Difference between bounds is only due to effective SNR



## Further Results and Future Research

- Inner bound can be improved: Optimization over curves, e.g. [Akyol-Vishwanatha-Rose-Ramstad IT'14]
- Outer bound for low-delay JSCC can be improved [Ziv-Zakai IT'73]
- High dimensional curves
- Other low-delay JSCC techniques: e.g., repetitive quantization [Kleiner-Rimoldi GLOBECOM'09]
  - Easy to generalize to higher dimensions
- Vector x, vector u, scalar y: Simple extension of scalar setting!
- Rate-matched case with vector y: "n: 1 JSCC" is needed
  - Switch roles between Transmitter and Receiver
  - Improves over [Freudenberg-Middleton-Solo AC'10]

