

Understanding Regression-adjusted Control Variate

Sobolev embedding and rare event

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<https://2prime.github.io/>

ML Nowadays

“entertainment, advising”

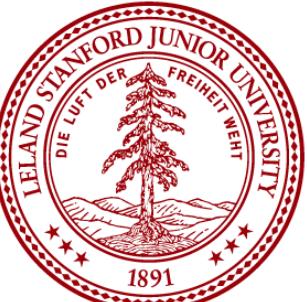


Midjourney

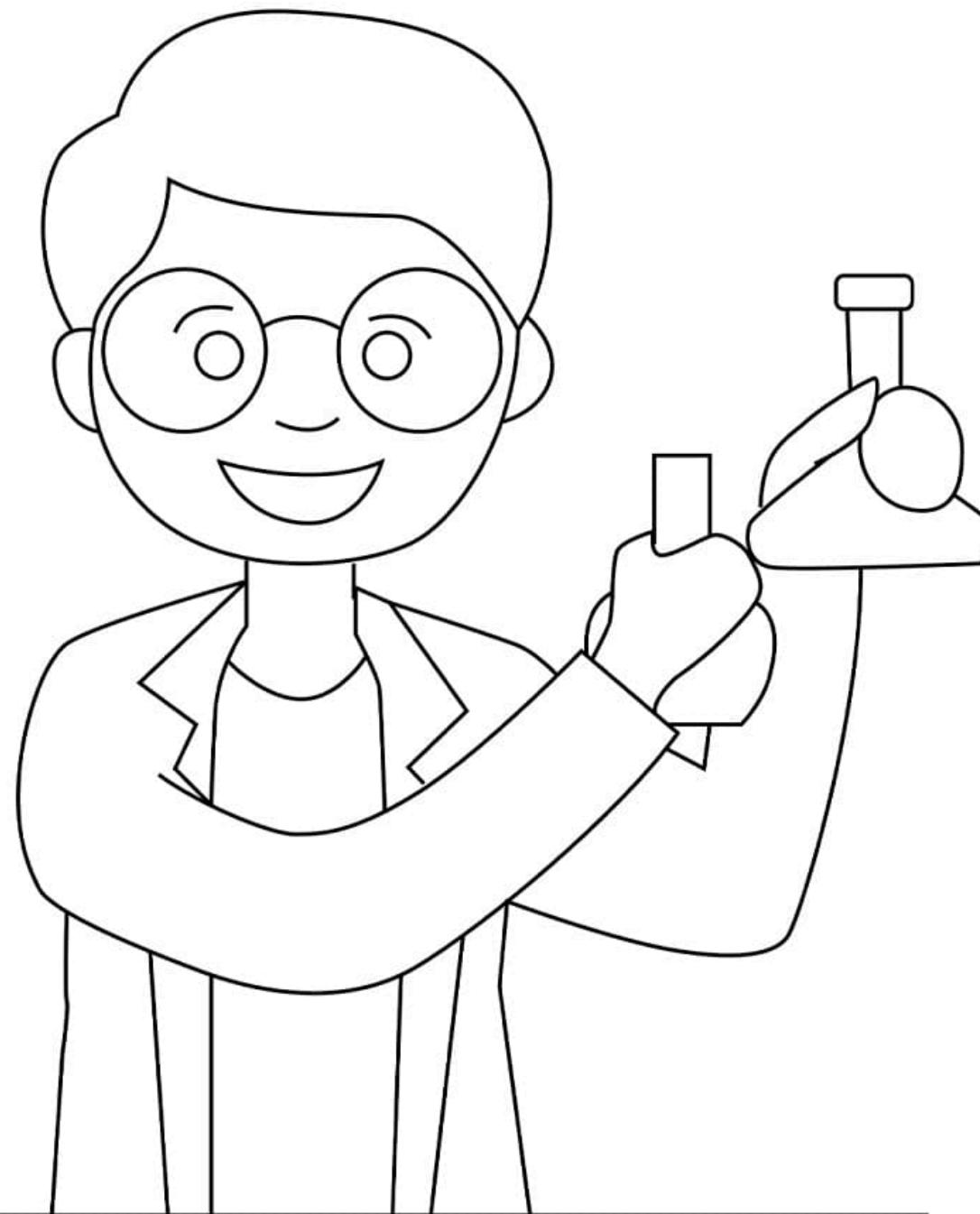


OpenAI
ChatGPT **4.0**

《science》



We want Guarantee!

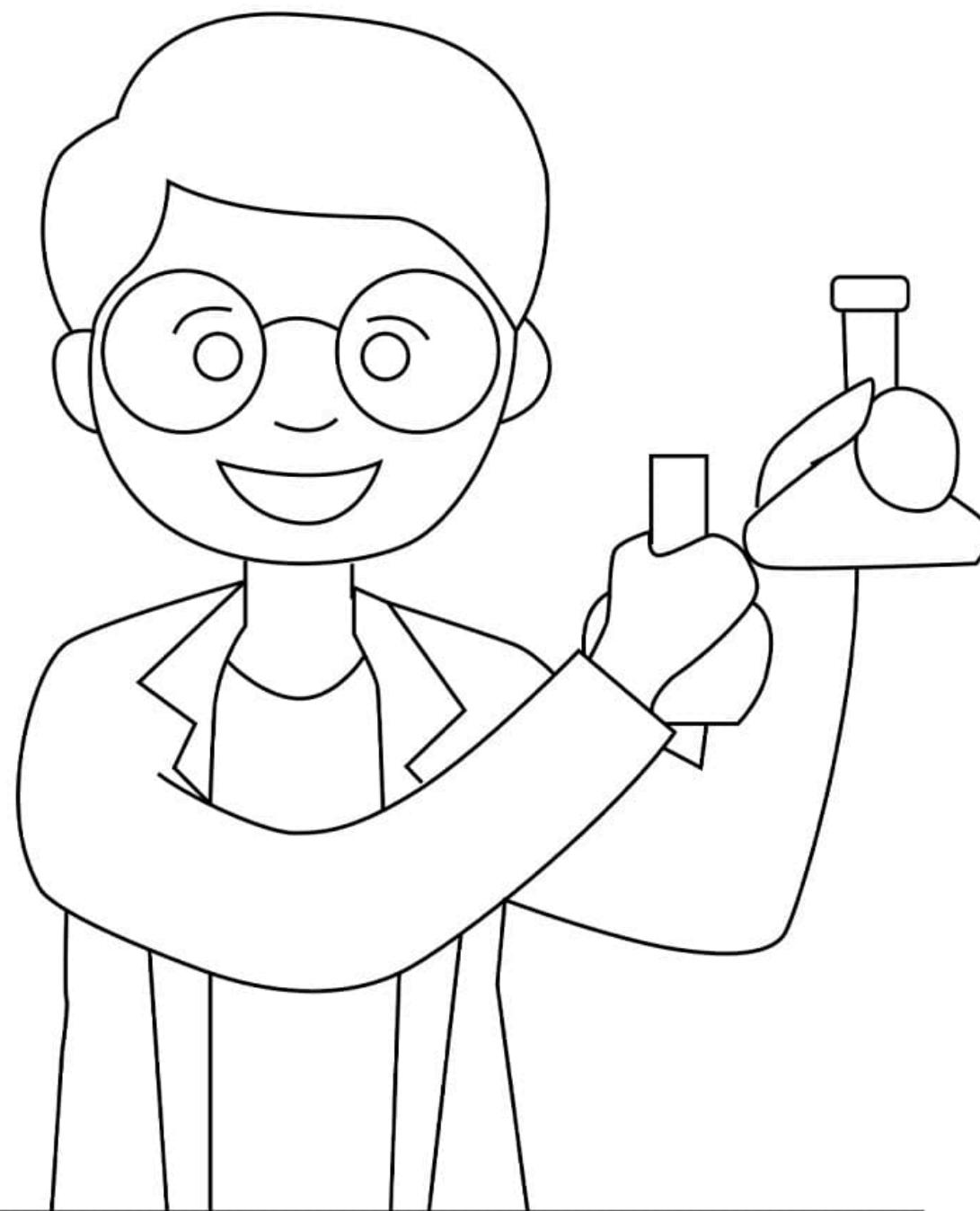


Theorem If you randomly collect
() data, then you can achieve
() accuracy with your AI!



We want Guarantee!

Big constant

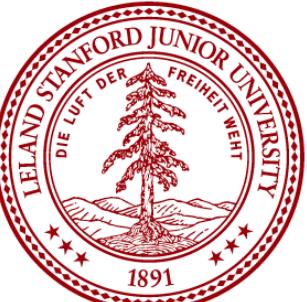


Theorem: If you randomly collect
() data, then you can achieve
() accuracy with your AI!

Assumption 1. Xxx
Assumption 2. Xxx
Assumption 3. Xxx
Assumption 4. Xxx



ML for Science nowadays



Debiasing ML for Science



You can prove theorem, but I still don't trust you!

Can we debias ML estimator or use it in an unbiased way?

Debiasing ML for Science

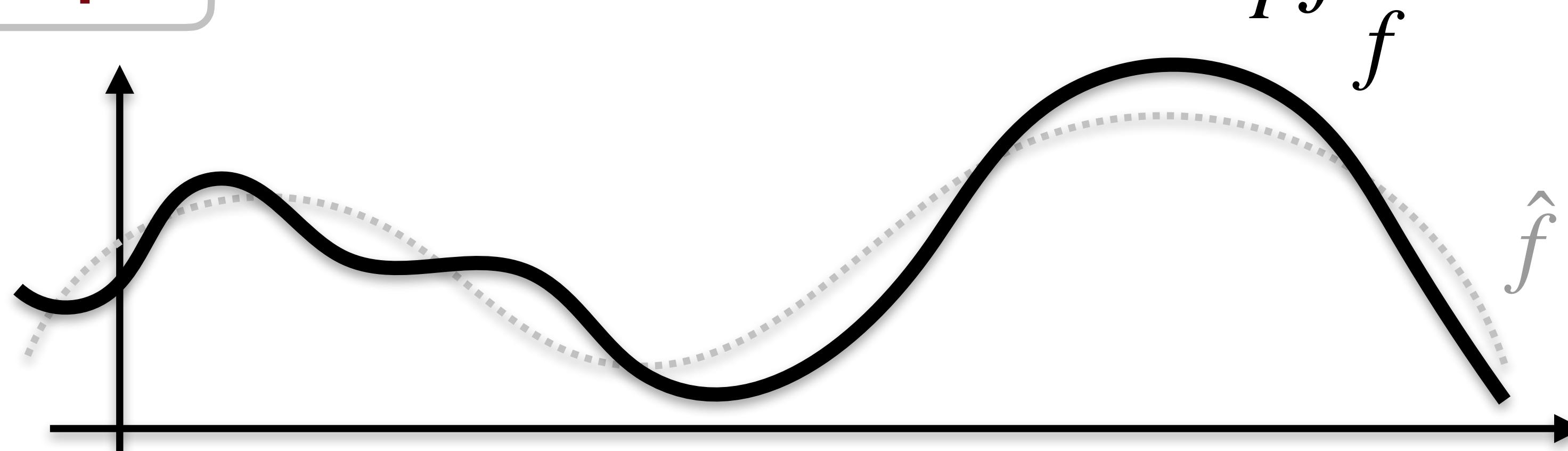
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Example

Monte Carlo Estimate $\mathbb{E}_P f$



Debiasing ML for Science

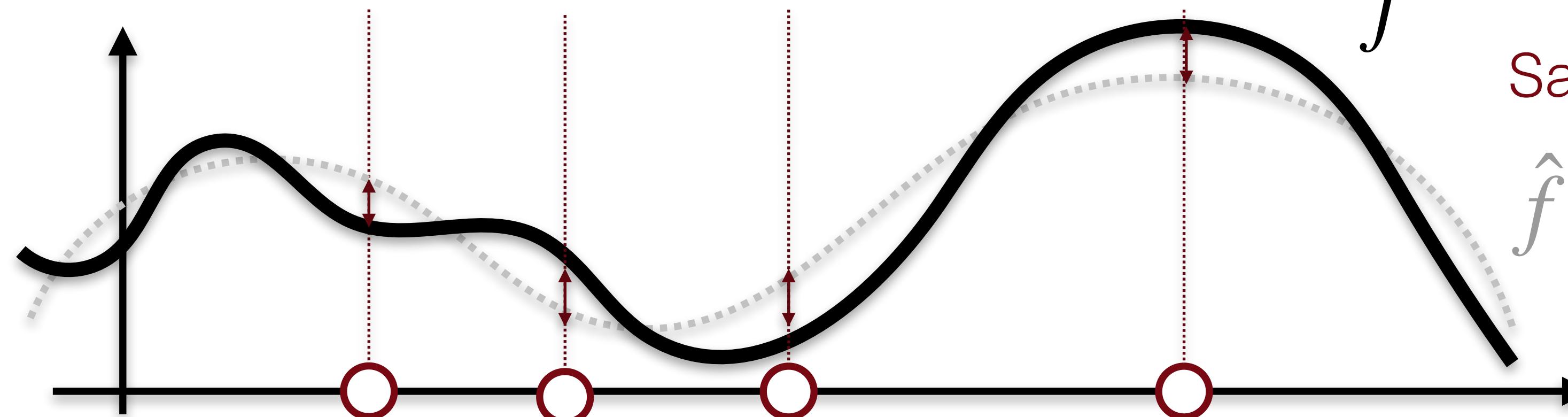
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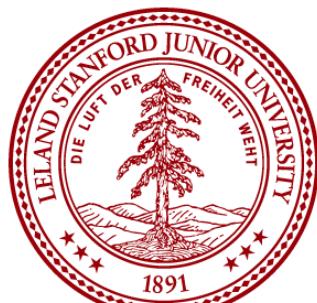
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Example

Monte Carlo Estimate $\mathbb{E}_P f$



Sample extra data to know $f - \hat{f}$



Debiasing ML for Science

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“Regression-adjusted control variate”

Can we debias ML estimator or use it in an unbiased way?

Example

Monte Carlo Estimate $\mathbb{E}_P f$

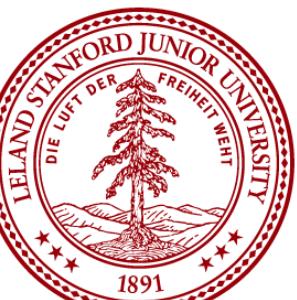
Step 1

Using half of the data to estimate \hat{f}

Step 2

$$\mathbb{E}_P f = \mathbb{E}_P(\hat{f}) + \mathbb{E}_P(f - \hat{f})$$

Low order term



“Modern” regression-adjusted cv

Trace estimation:

Hutch++ Lin 17 Numerische Mathematik Mewyer-Musco-Musco-Woodruff 20

Dimension Reduction:

Sobczyk and Luisier Neurips 22

Conformal Prediction:

Conformalized quantile regression Romano-Patterson-Candes Neurips 19

Gradient Estimation

Shi-Zhou-Hwang-Tisias-Mackey Neurips 22 outstanding paper

Causal Inference:

Double Robust estimation



Debiasing ML for Science



Is this algorithm statistical optimal?

When this improves MC estimator?

Can we debias ML estimator or use it in an unbiased way?

Example

Monte Carlo Estimate $\mathbb{E}_P f$

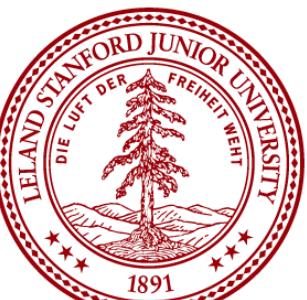
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Low order term



Understanding this statistically...



Is this algorithm statistical optimal?

Why consider q -th moment?

When this improves MC estimator?

Can we debias ML estimator or use it in an unbiased way?

Example

Monte Carlo Estimate $\mathbb{E}_P f$ $\mathbb{E}_P f^q, f \in W^{s,p}$

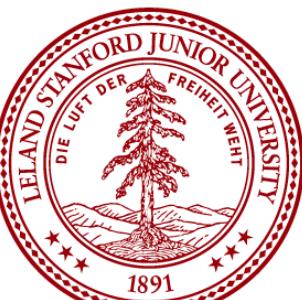
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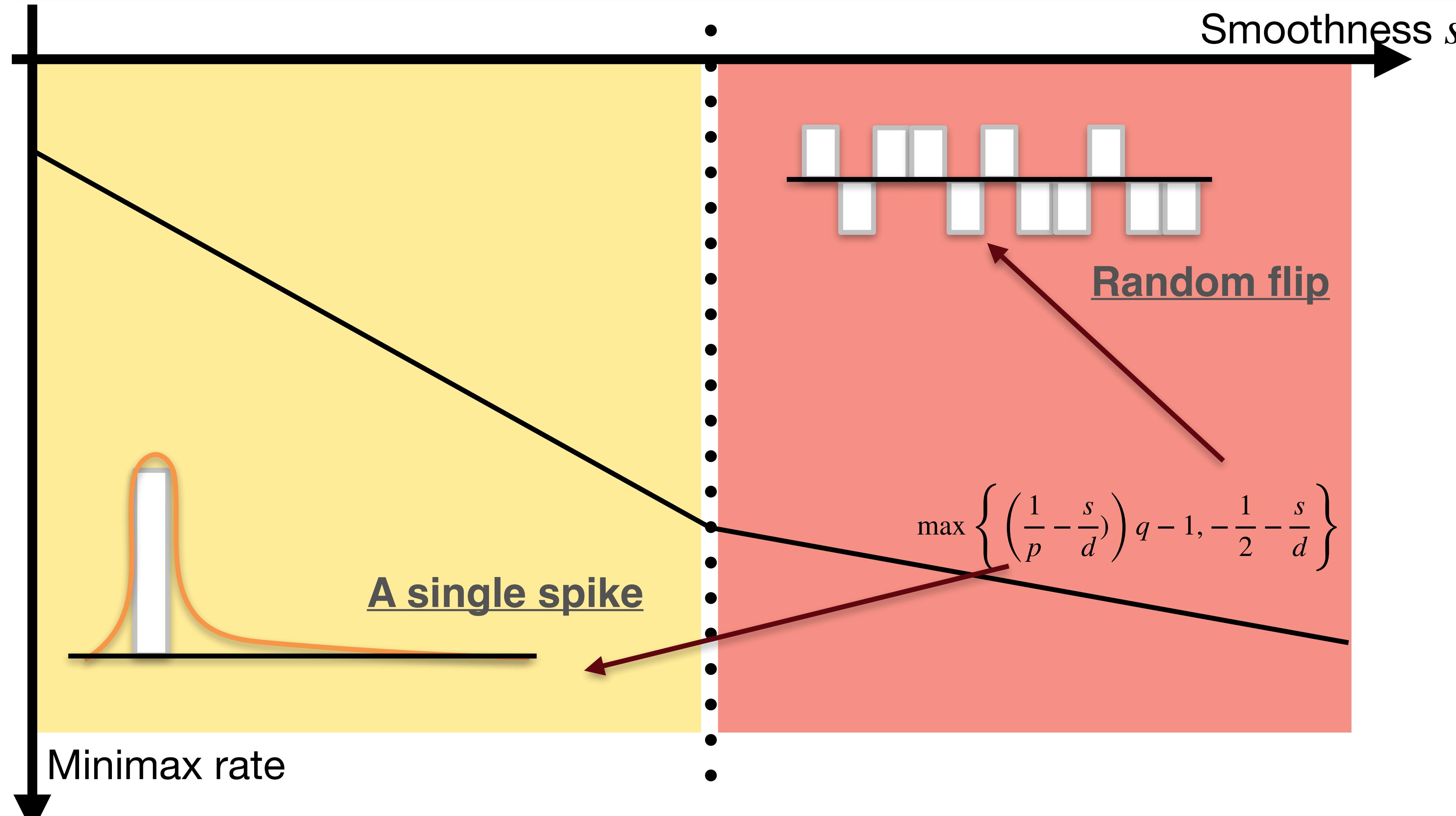
Step 2

$$\mathbb{E}_P f^q = \mathbb{E}_P (\hat{f})^q + \mathbb{E}_P (f - \hat{f})^q$$

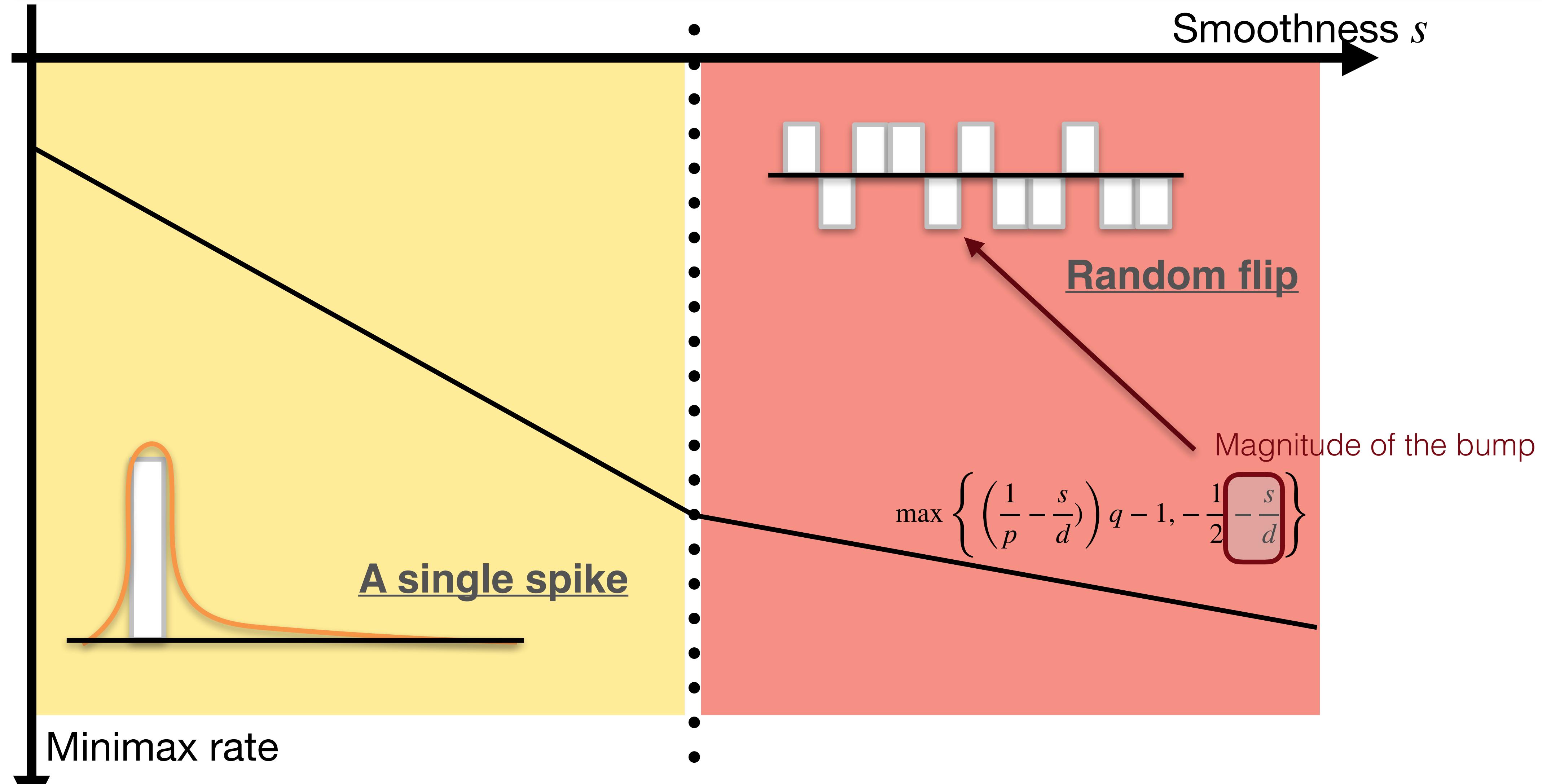
Low order term



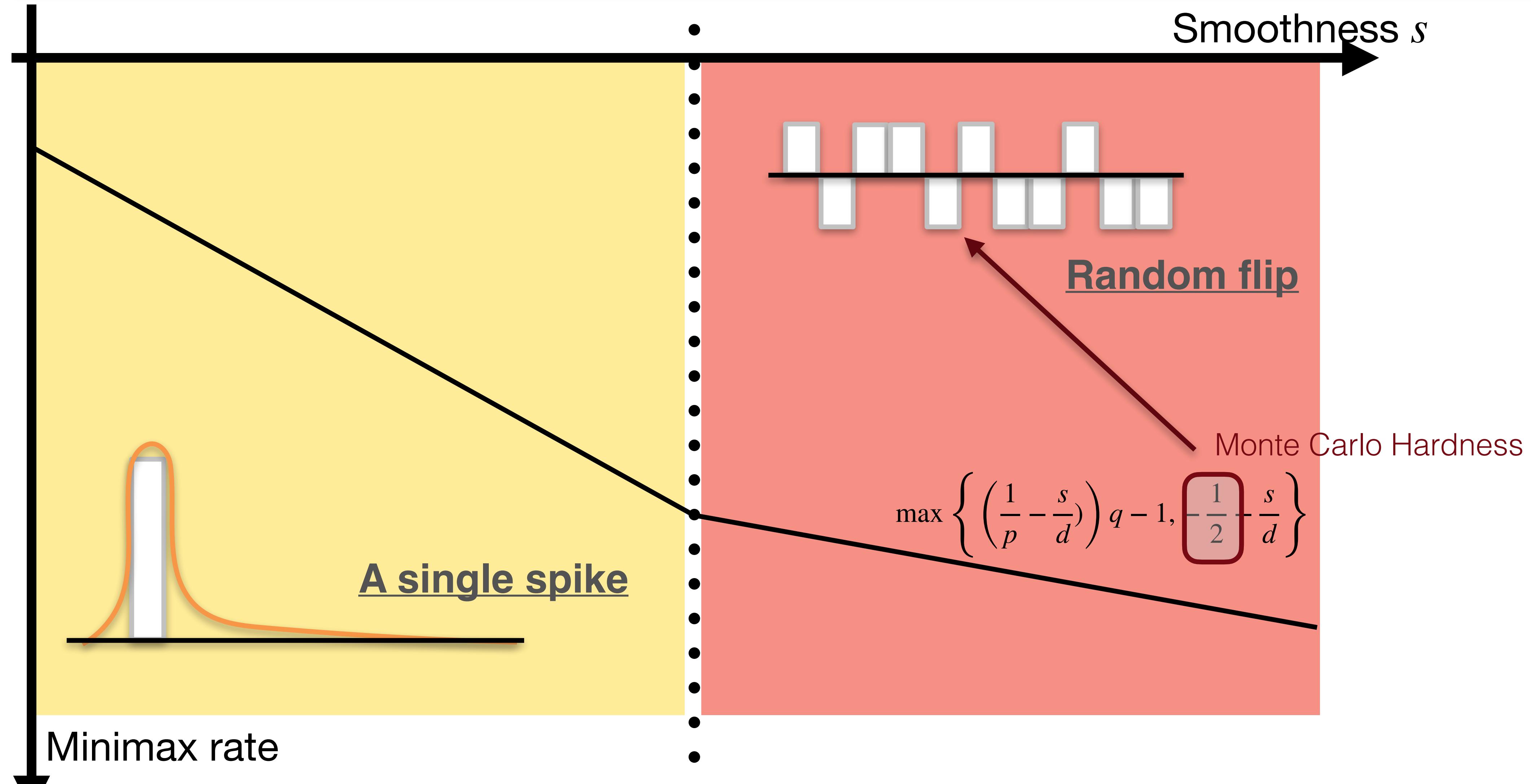
Setting the information theoretical limit



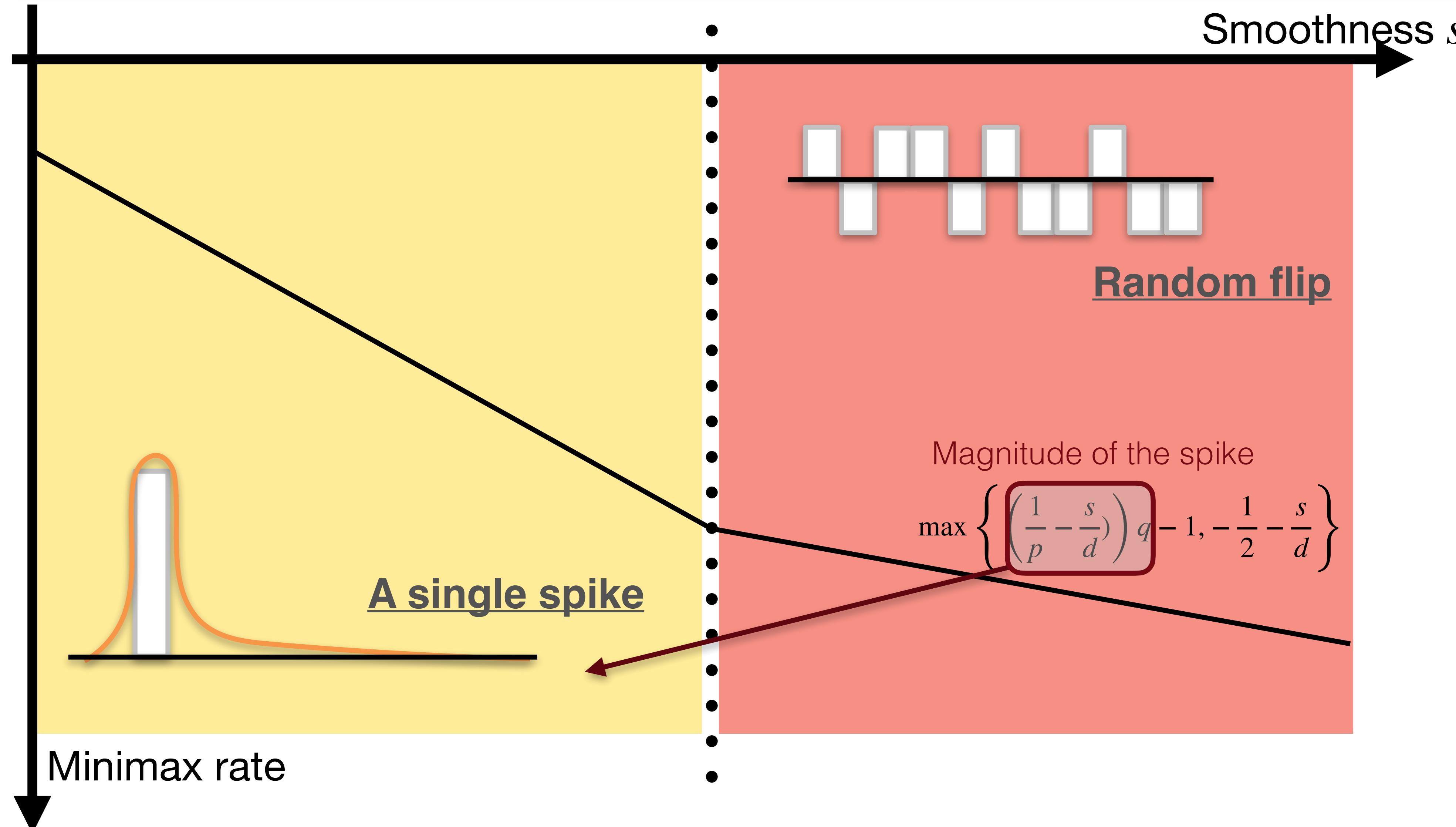
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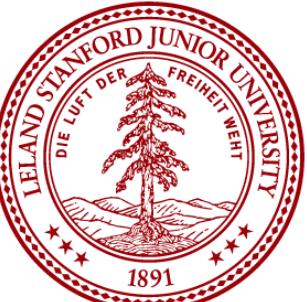
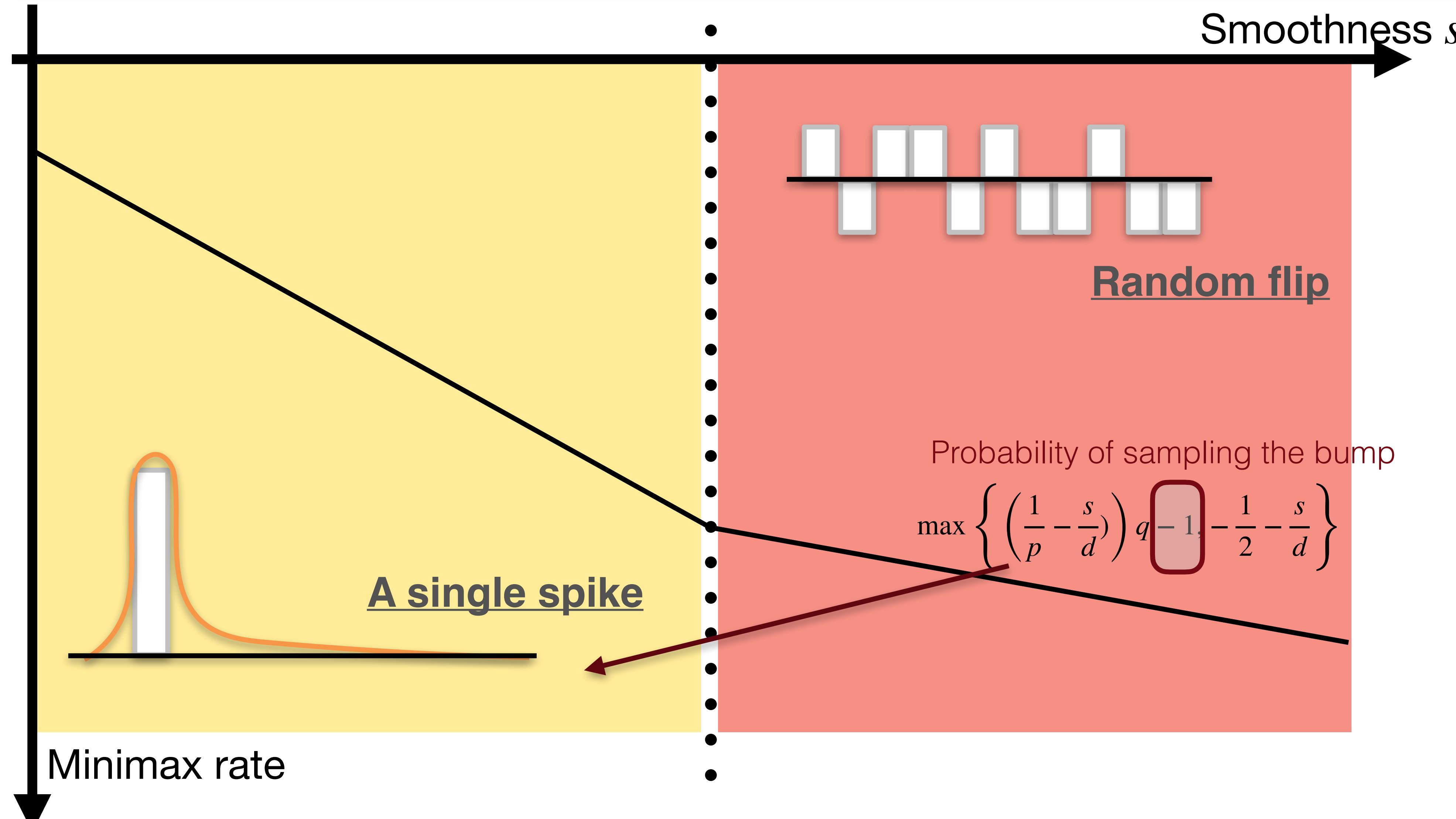
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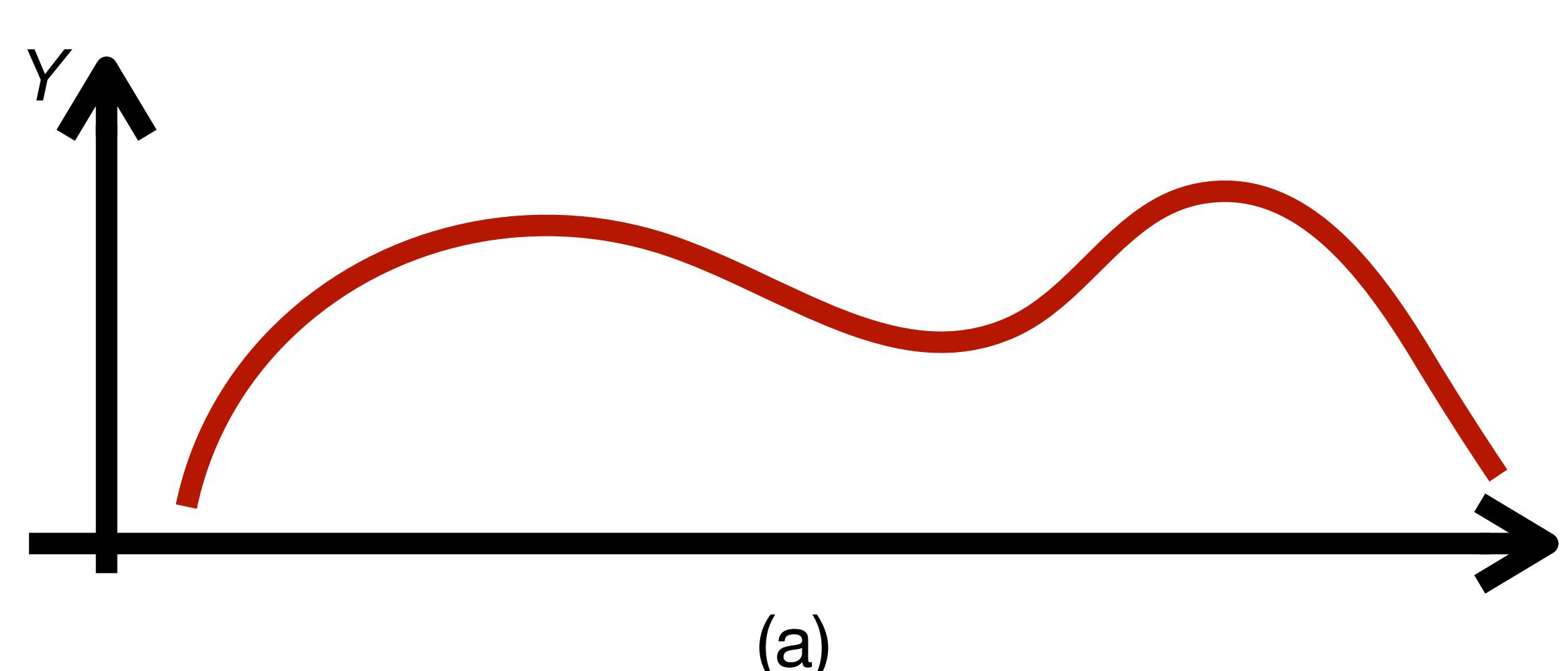
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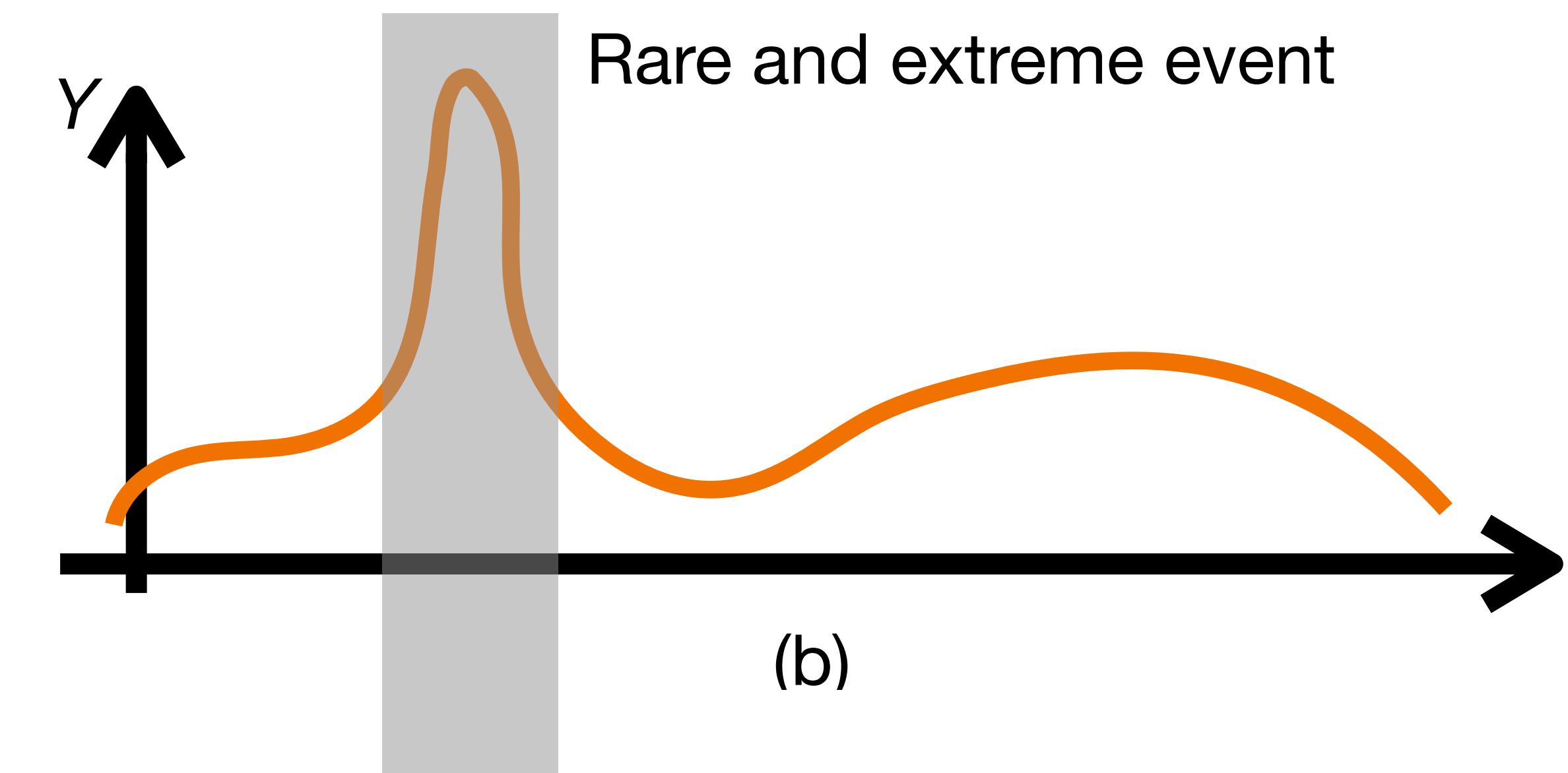
Setting the information theoretical limit



Rare Event and Smoothness...



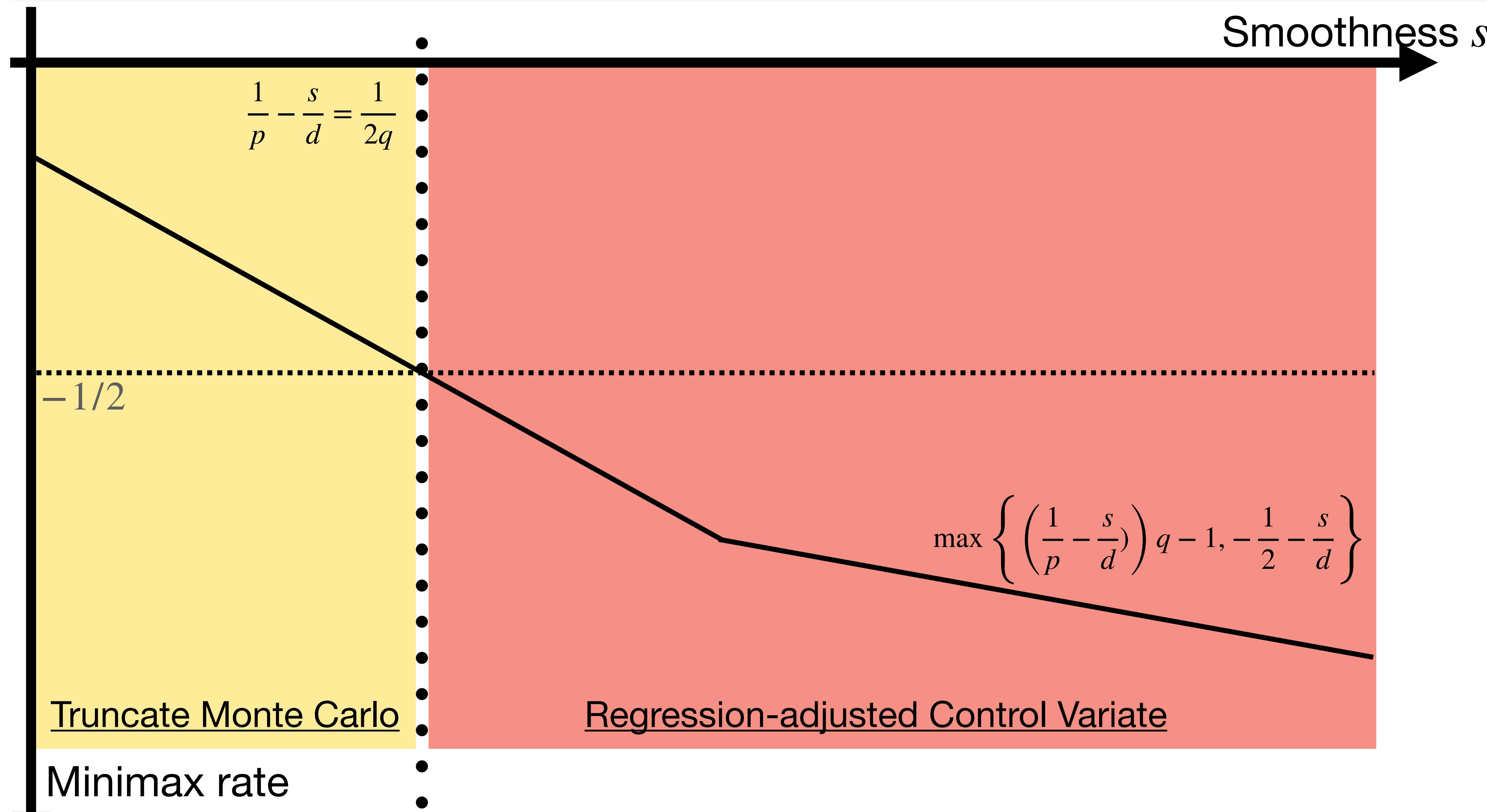
(a)



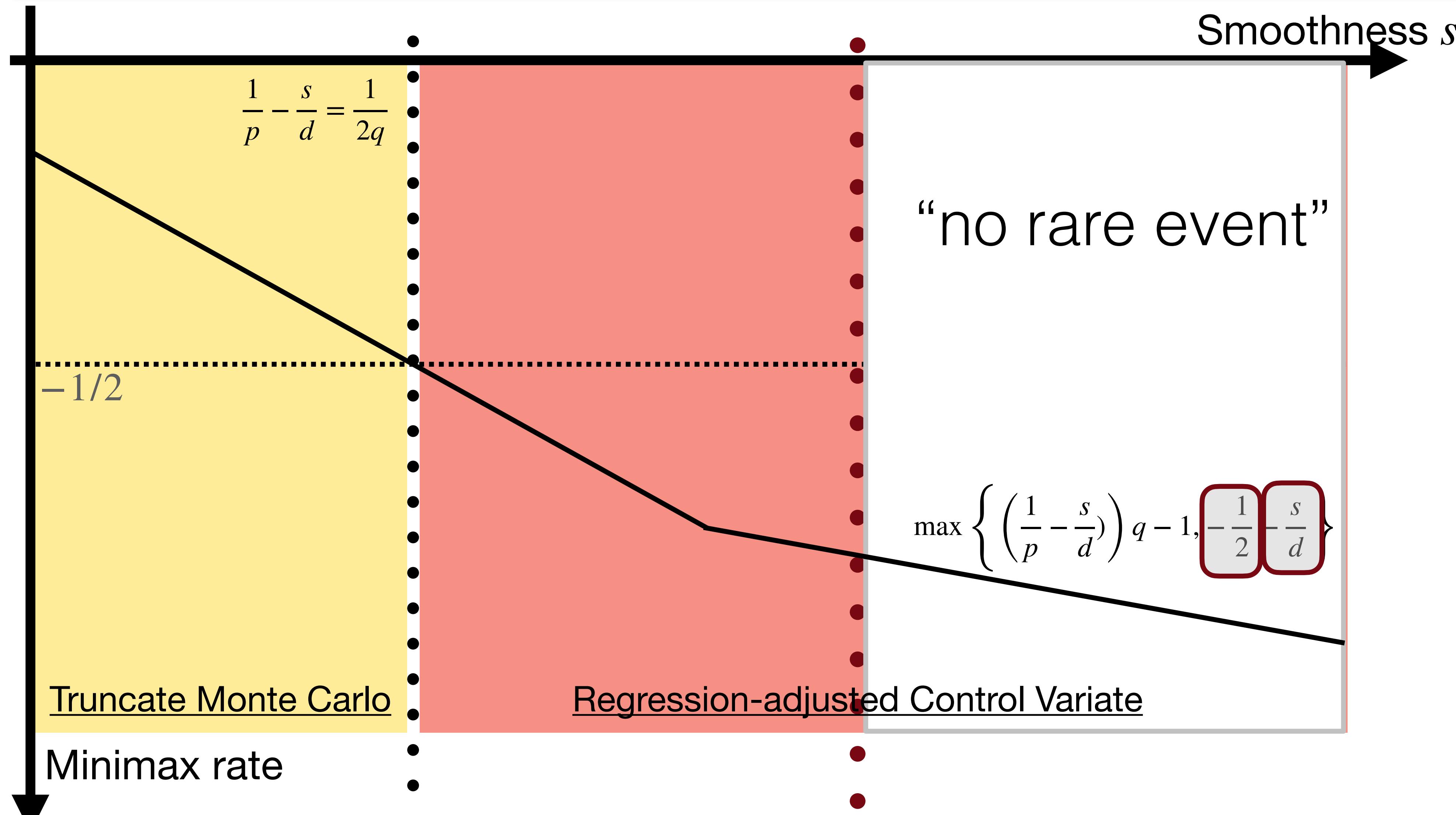
(b)

Rare and extreme event

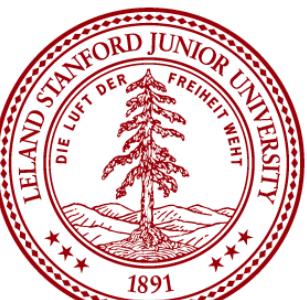
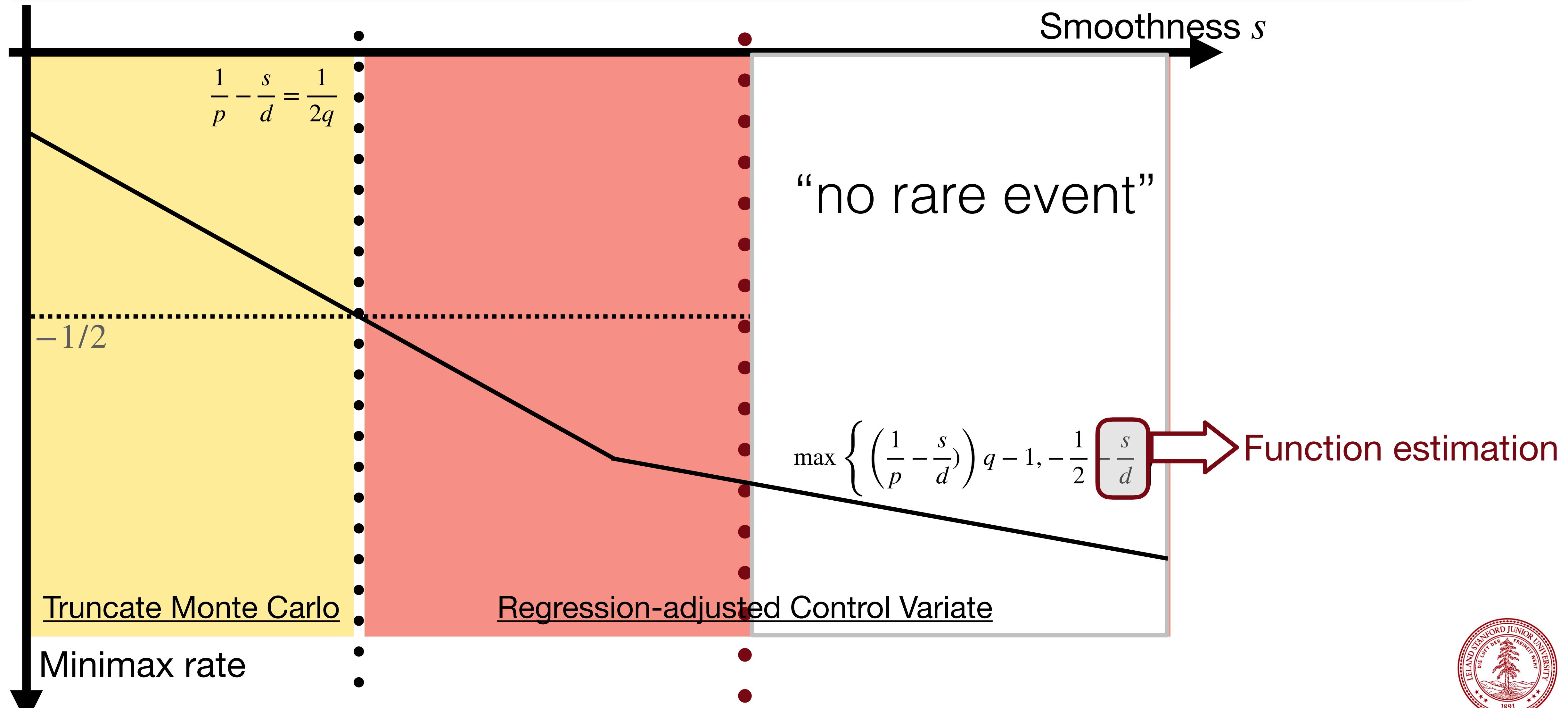
When the control variate helps



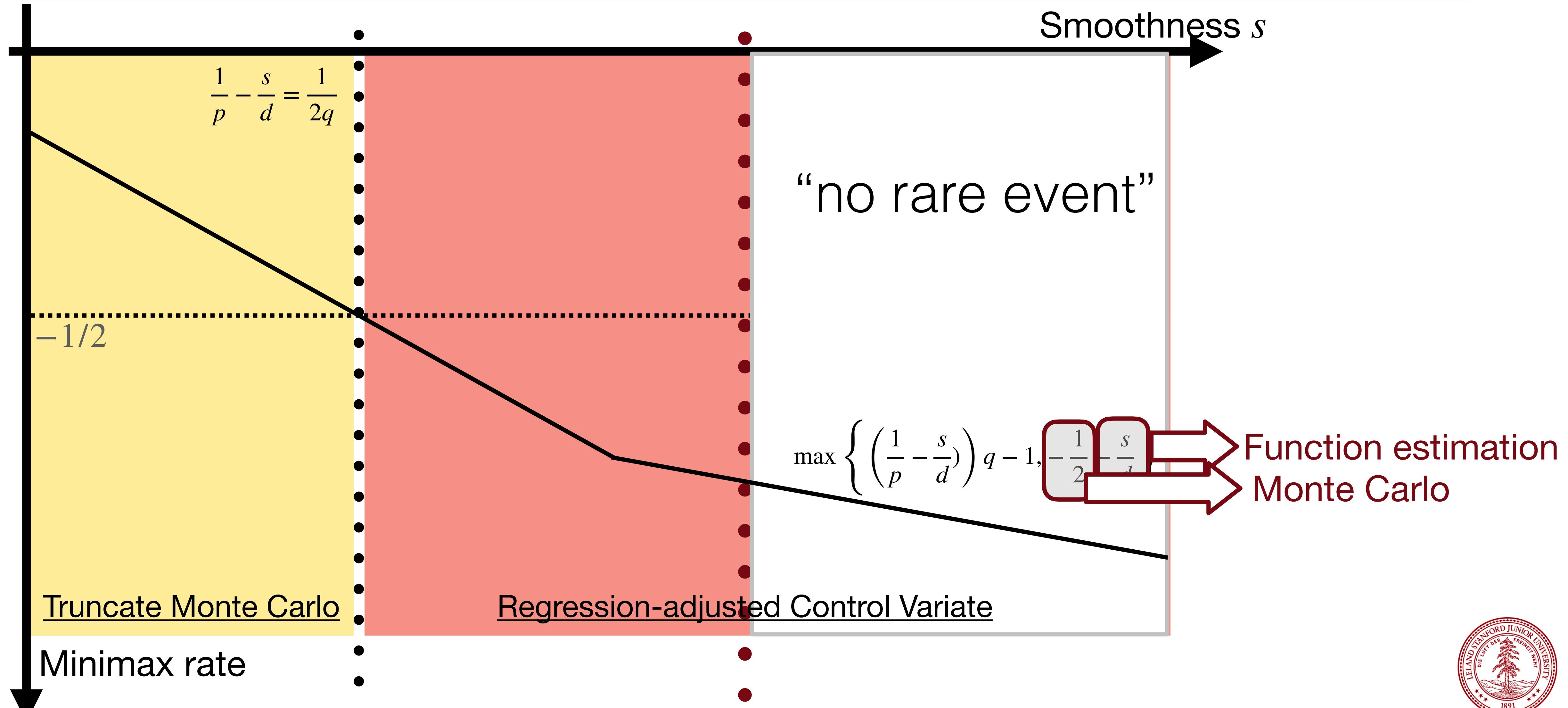
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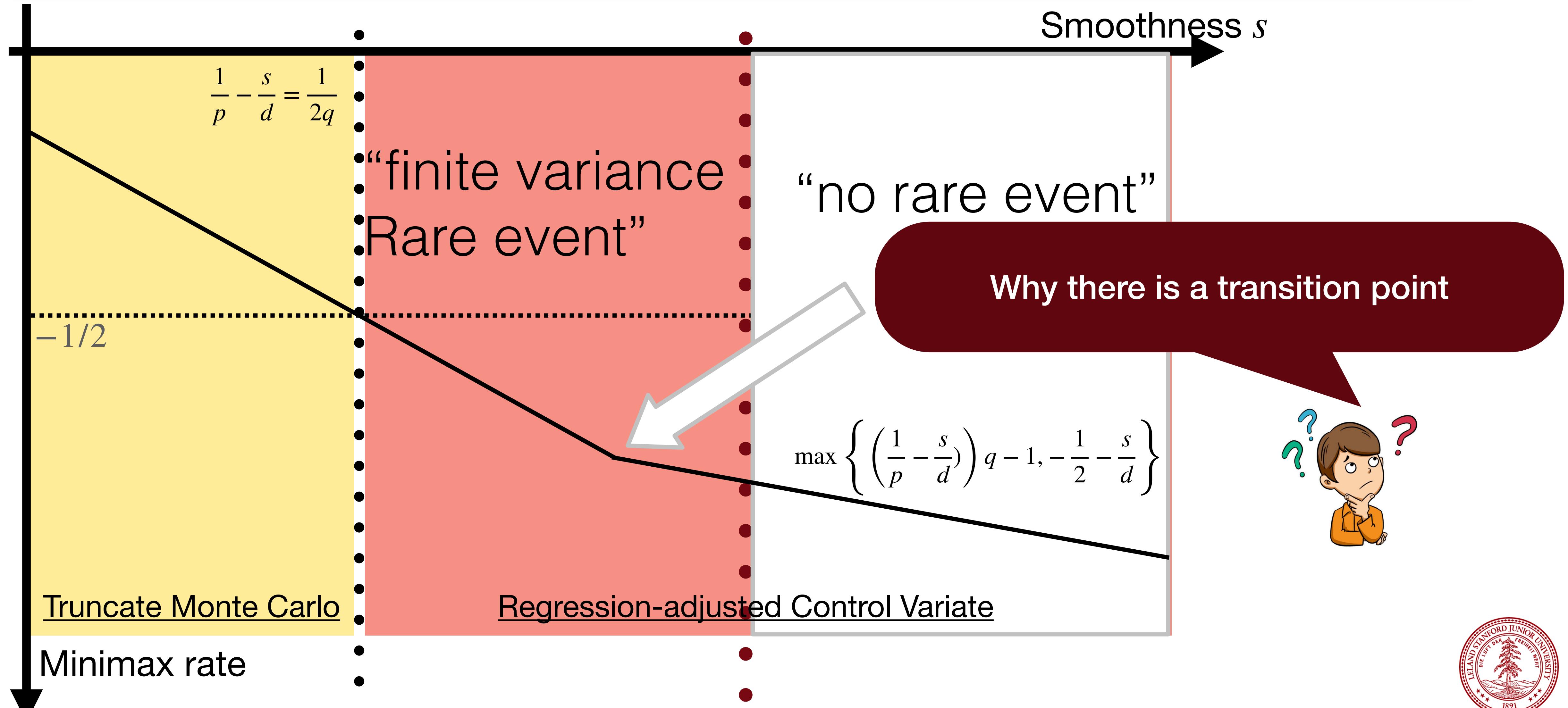
When the control variate helps



When the control variate helps



When the control variate helps



Semi-parametric efficiency...

Example

Monte Carlo Estimate $\mathbb{E}_P f$ $\mathbb{E}_P f^q, f \in W^{s,p}$

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$$\mathbb{E}_P f^q = \mathbb{E}_P (\hat{f})^q + \mathbb{E}_P (f - \hat{f})^q$$

Low order term

$$f^{q-1}(f - \hat{f}) + (f - \hat{f})^q$$

“influence function” (gradient)



Semi-parametric efficiency...

Example

Monte Carlo Estimate $\mathbb{E}_P f = \mathbb{E}_P f^q, f \in W^{s,p}$

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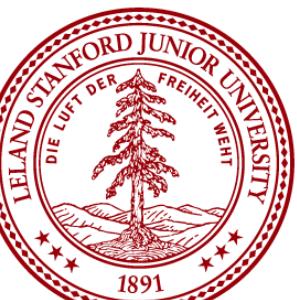
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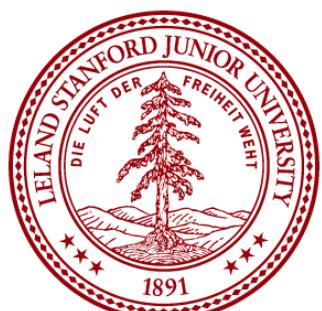
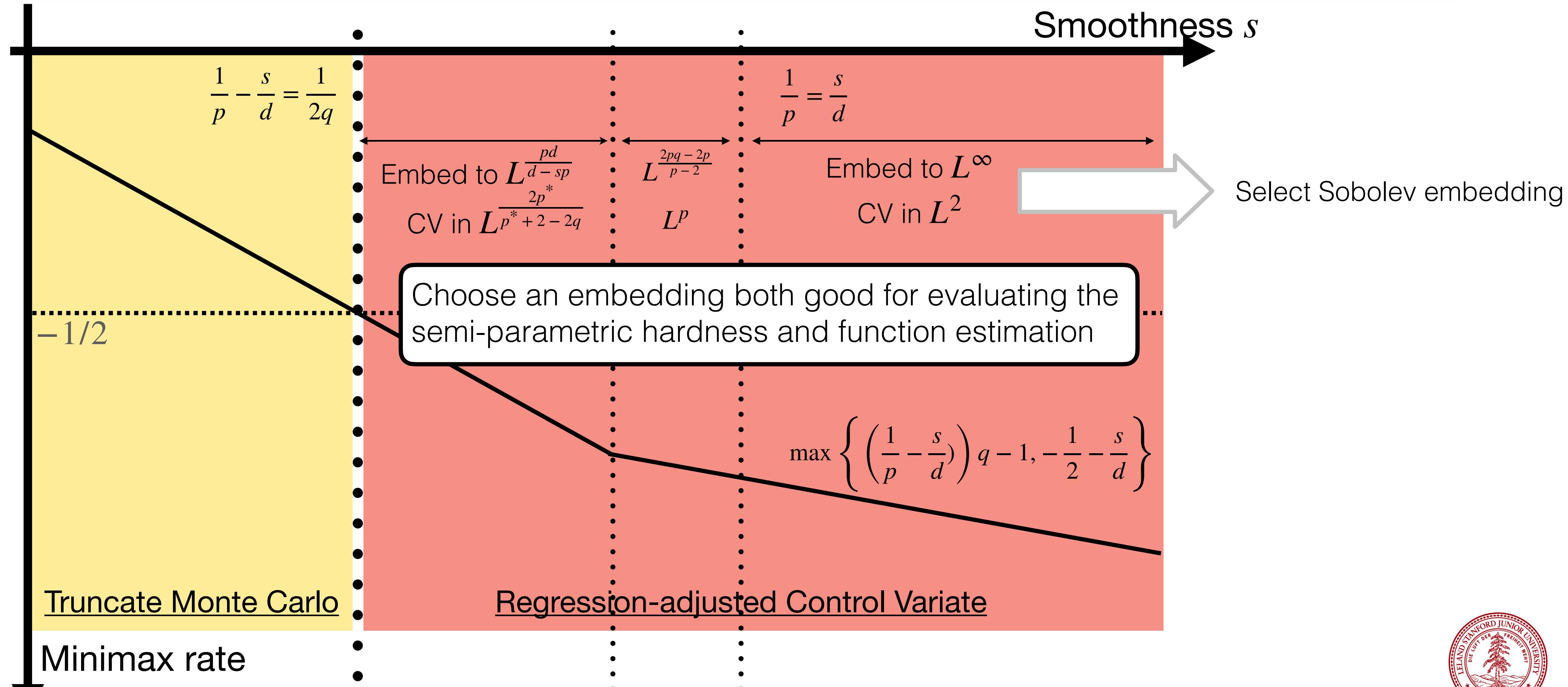
“influence function” (gradient)

Embed f^{q-1} and $f - \hat{f}$ into “dual” space

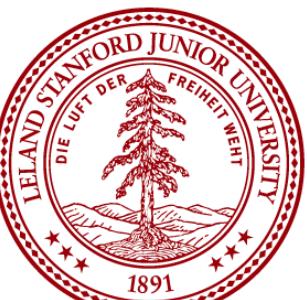
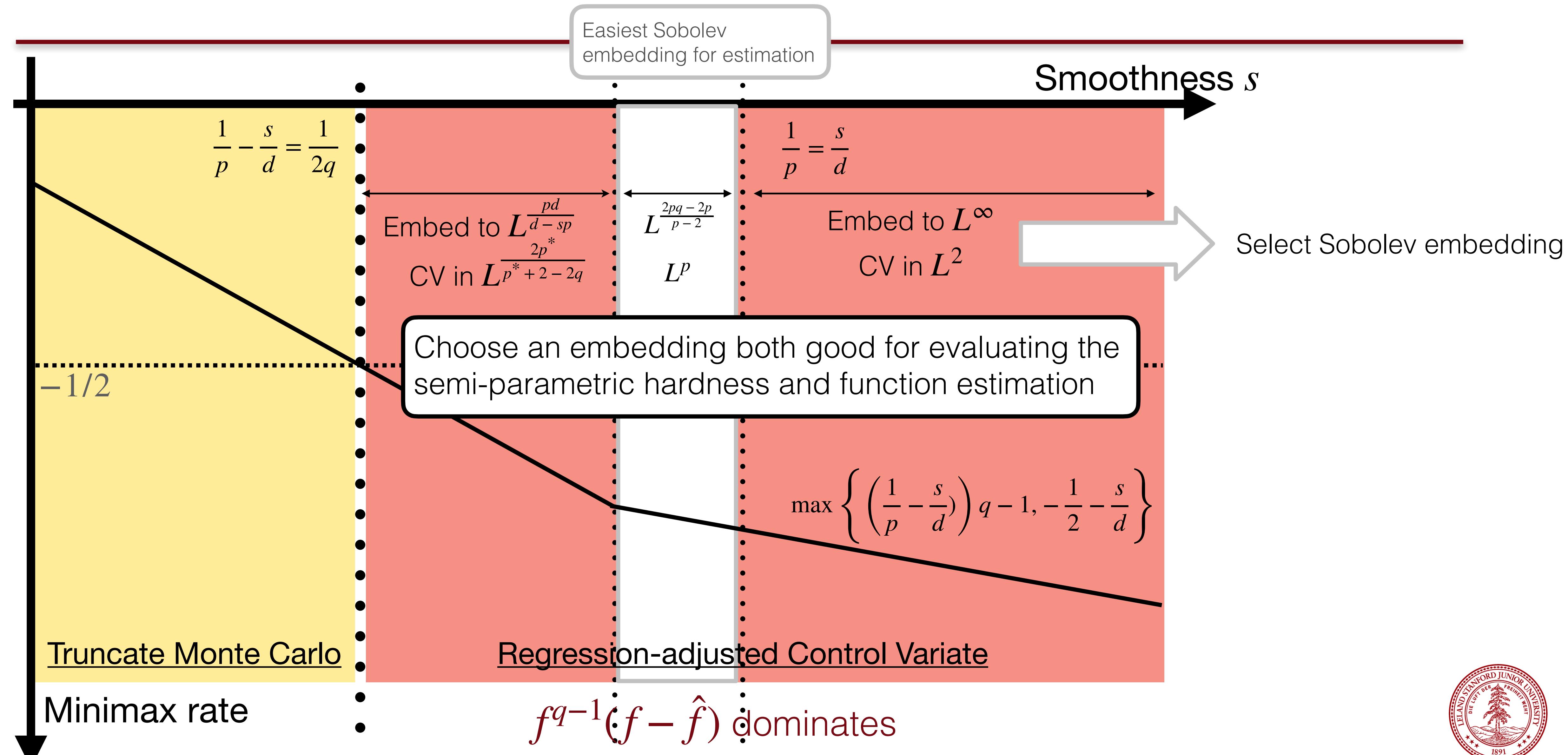
How to select the sobolev
emebedding



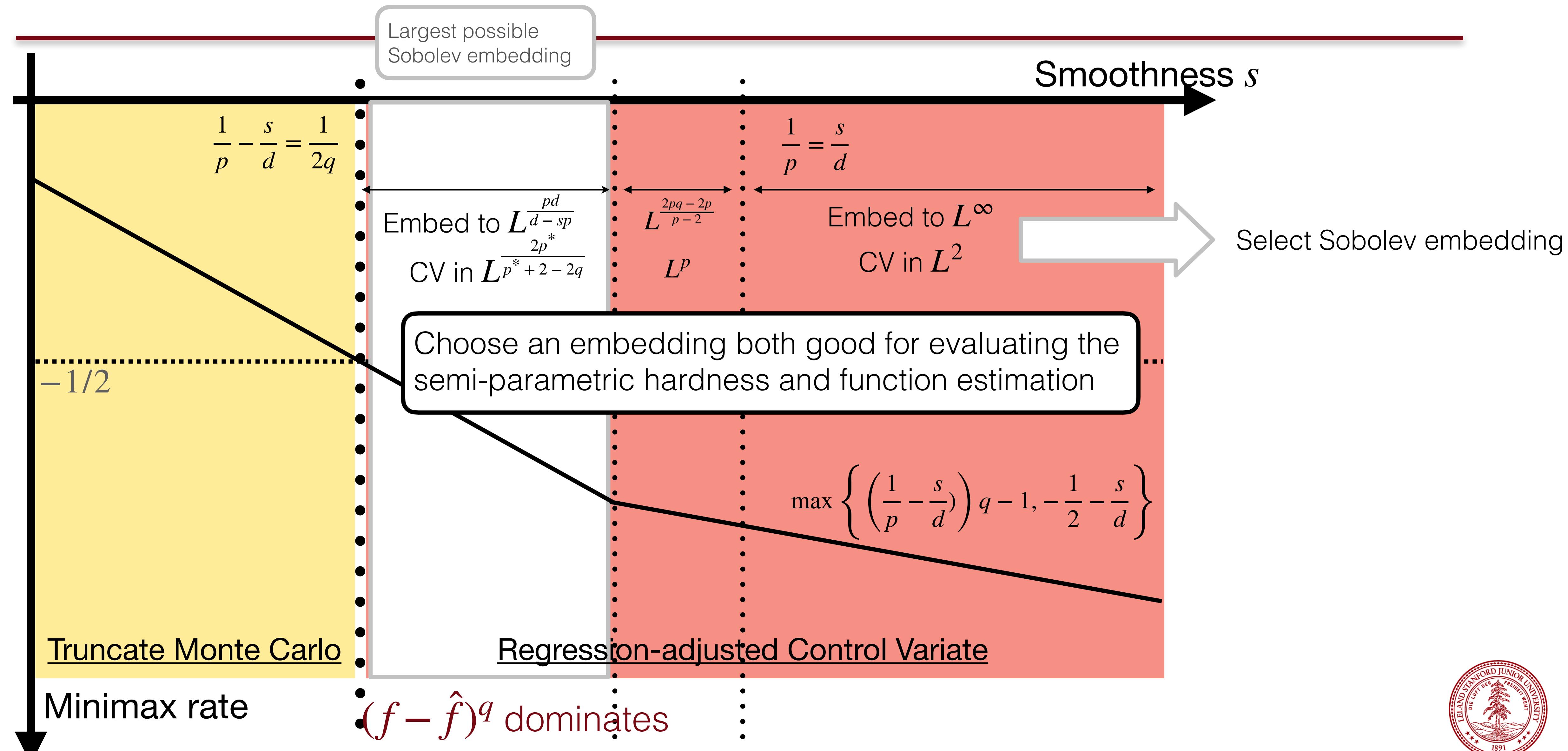
Tricky part of the Proof:select embedding



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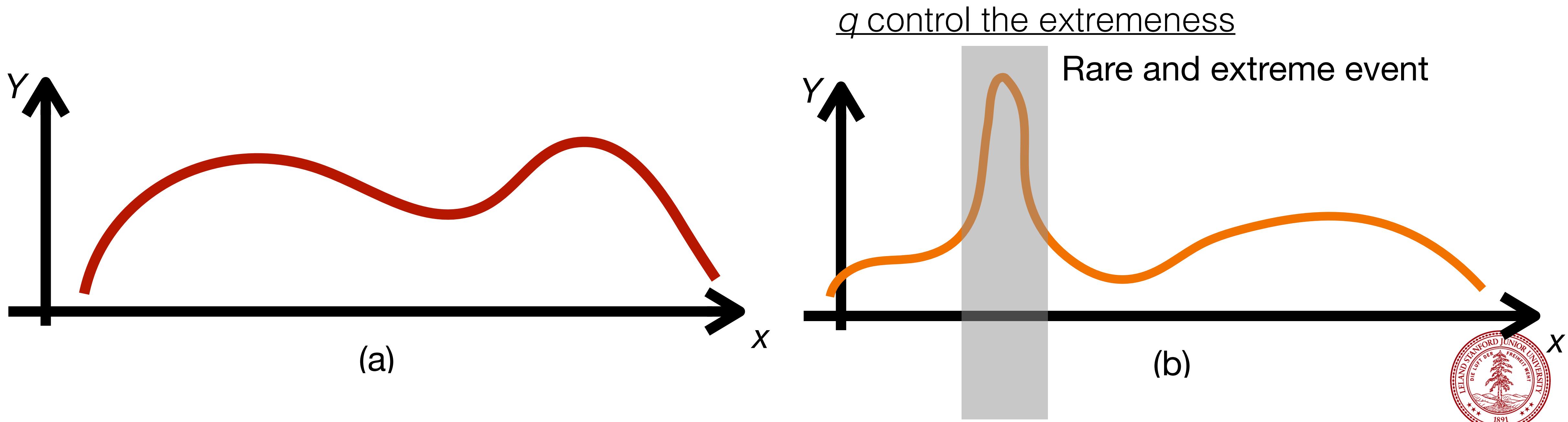


Tricky part of the Proof:select embedding



Take home message

- a) Statistical optimal regression is the optimal control variate
- b) It helps only if there isn't a hard to simulate (infinite variance)
Rare and extreme event





Contact: yplu@stanford.edu