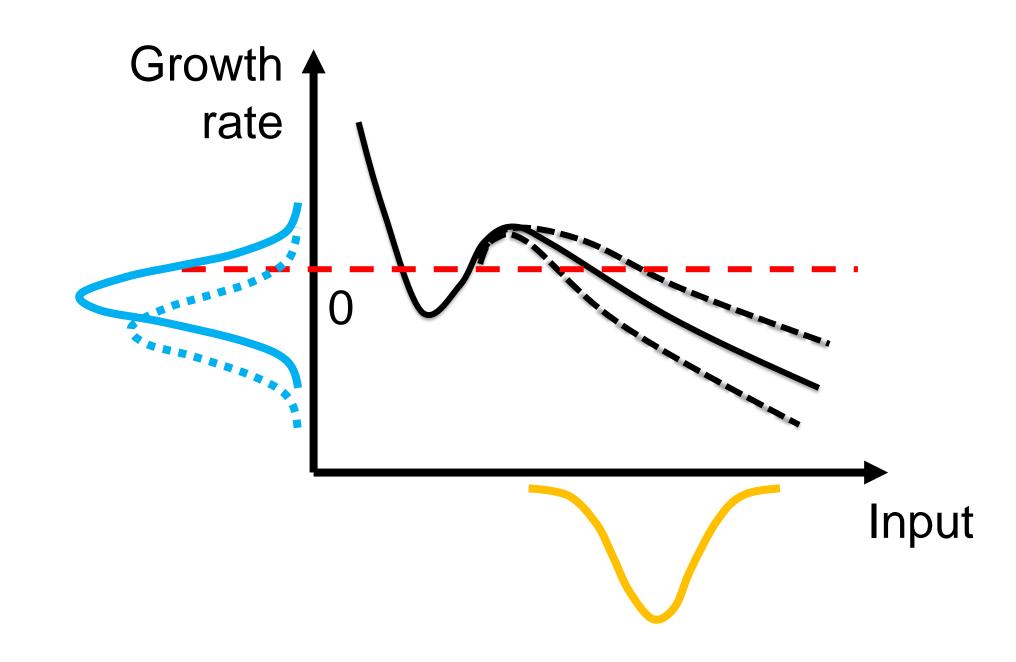
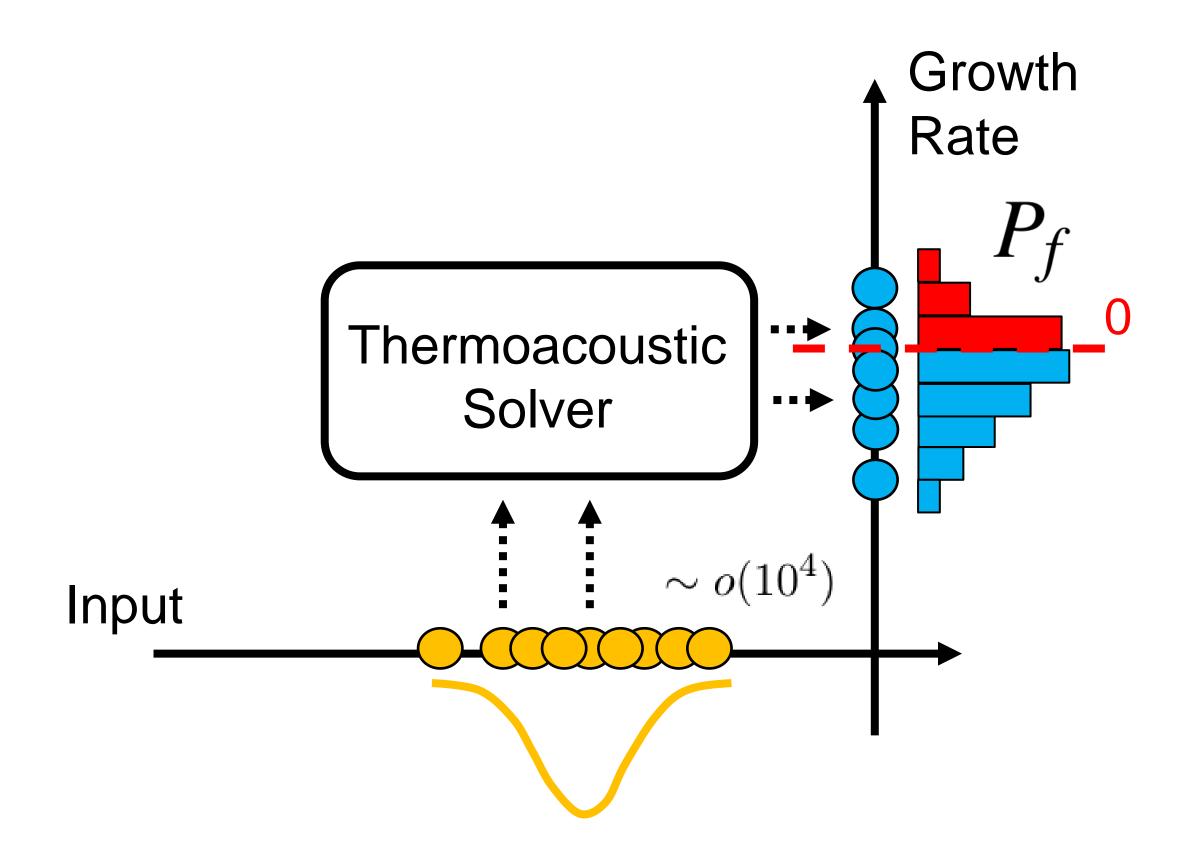
Reliable calculation of thermoacoustic instability risk using an imperfect surrogate model

S. Guo, C. Silva, W. Polifke

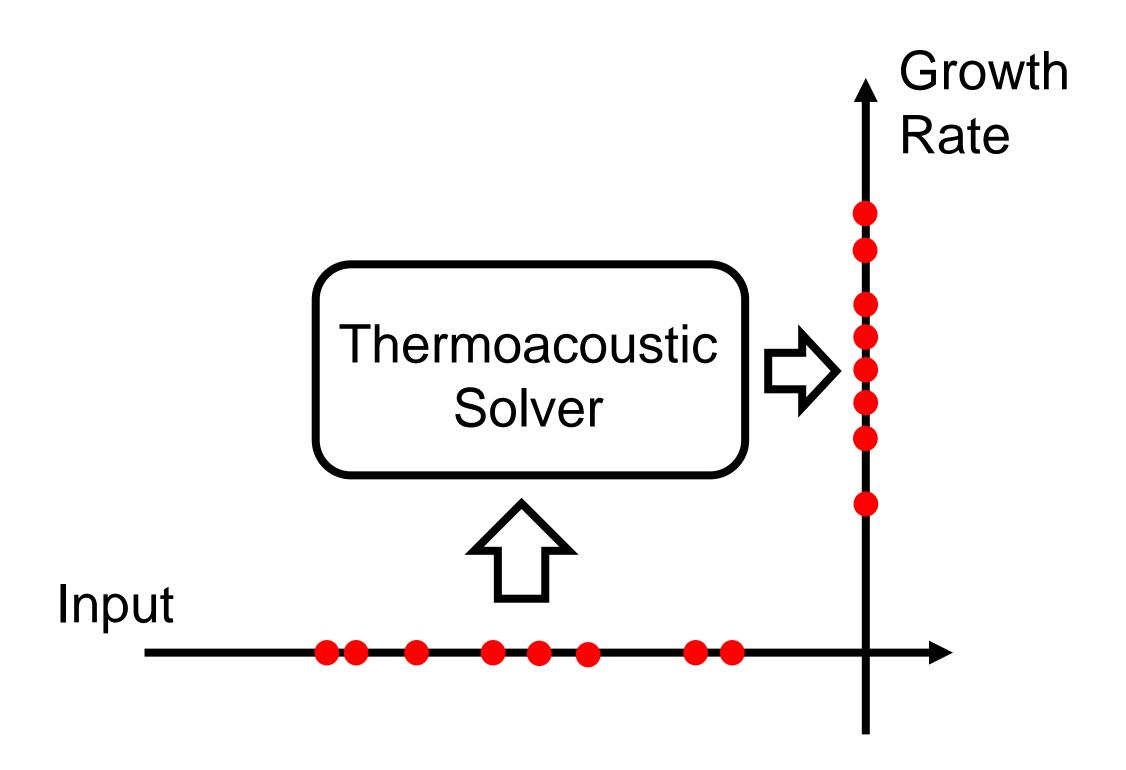


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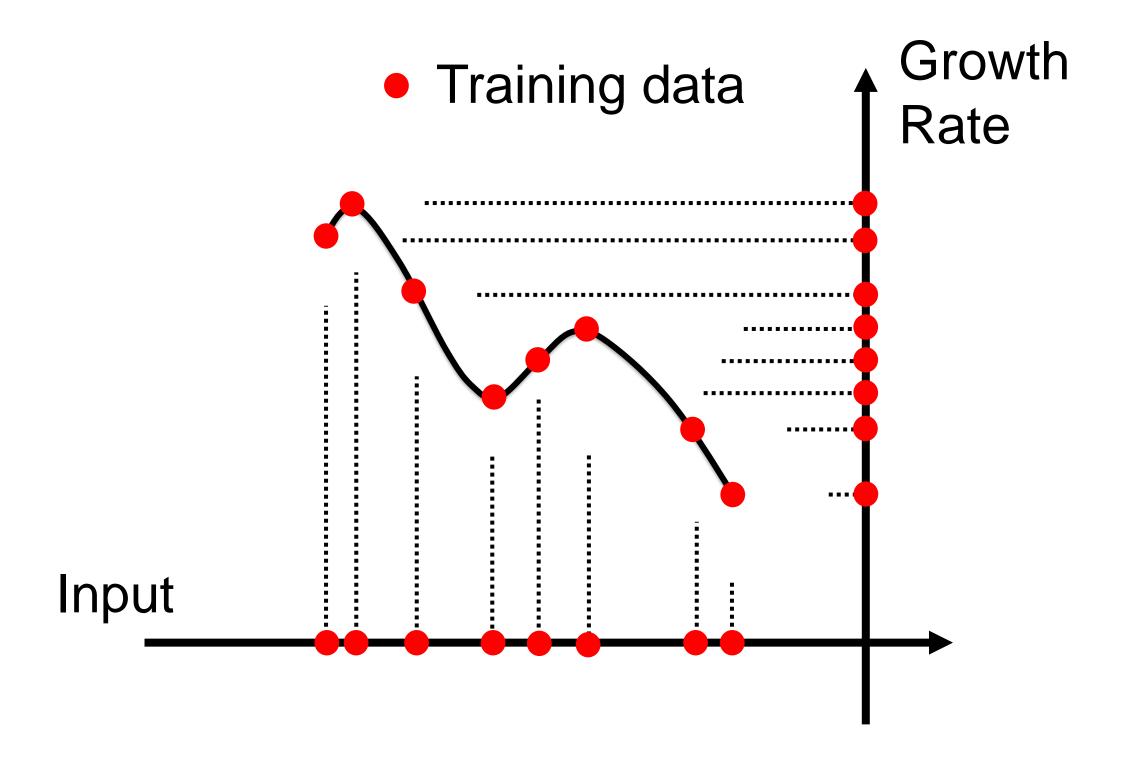




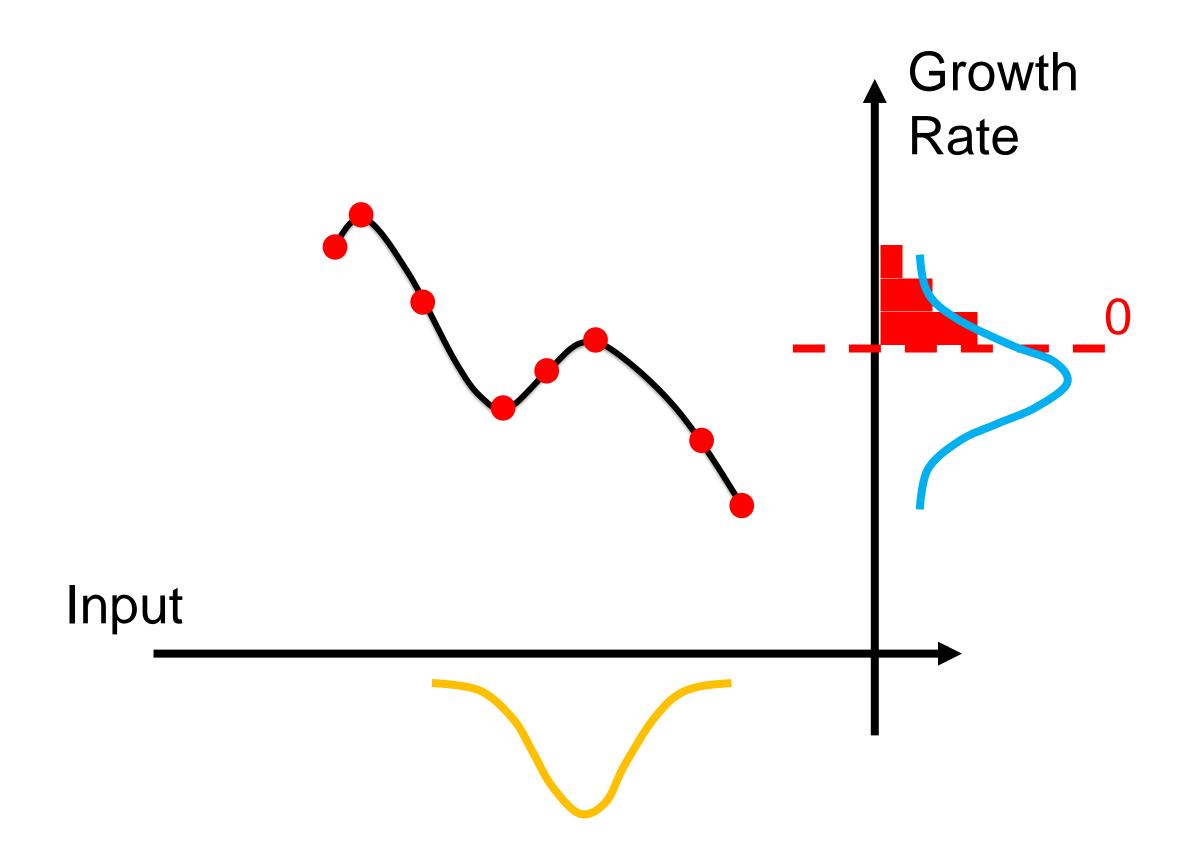




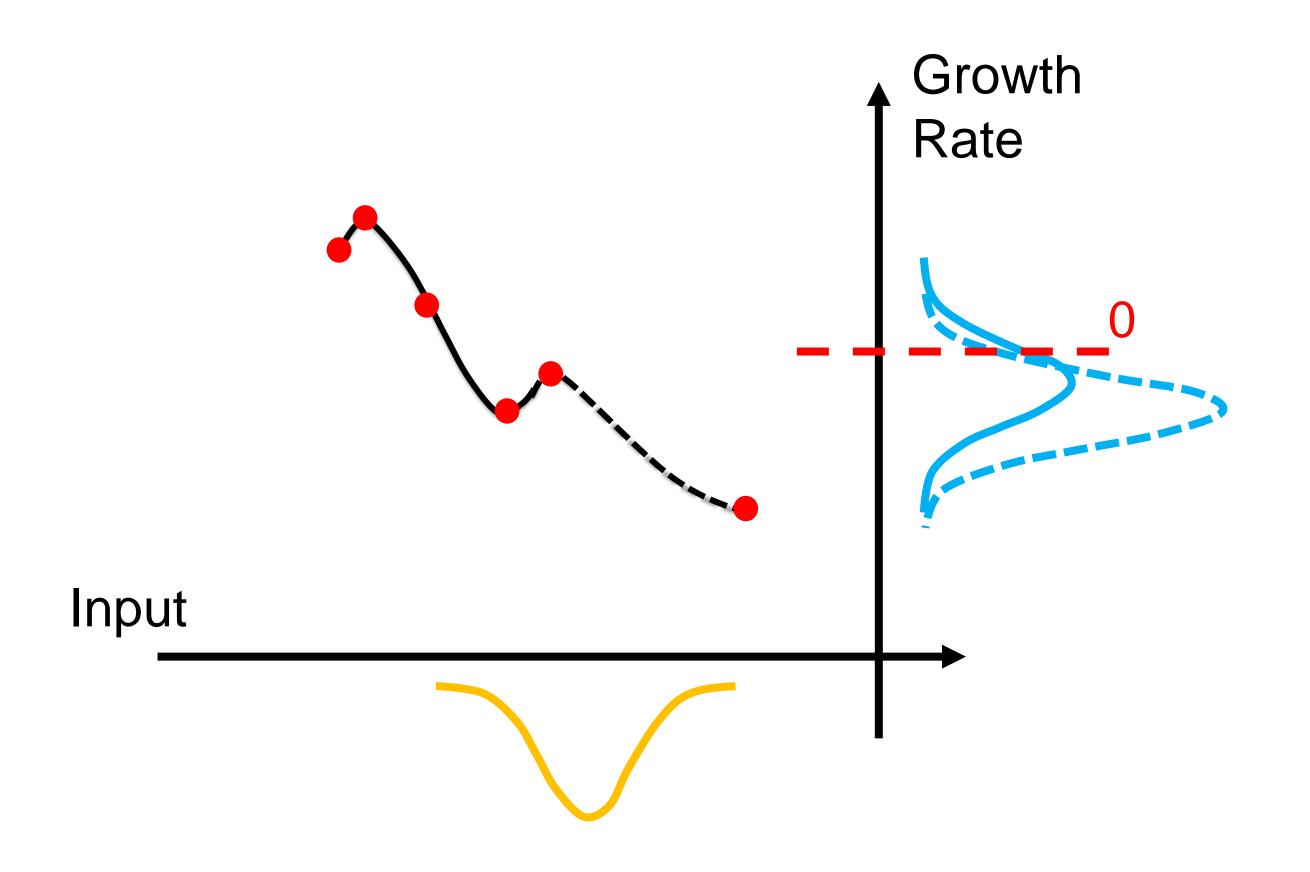








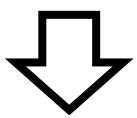




Limited training data



Uncertain surrogate model

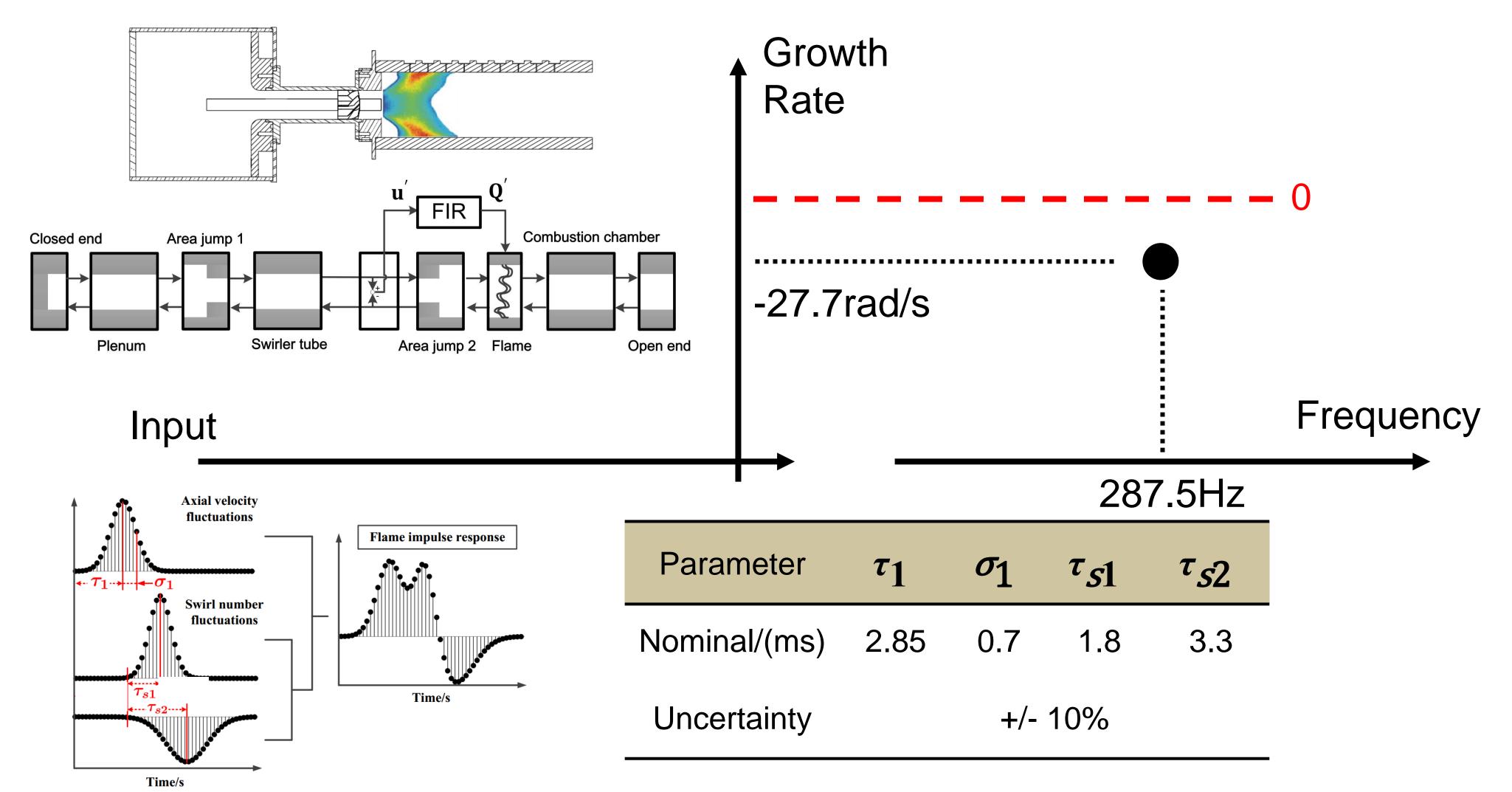


Uncertain risk calculation

Presentation overview

- Motivation
- Case study settings
- ☐ Quantifying the variation of risk calculation
- ☐ Reducing the variation of risk calculation
- Conclusions

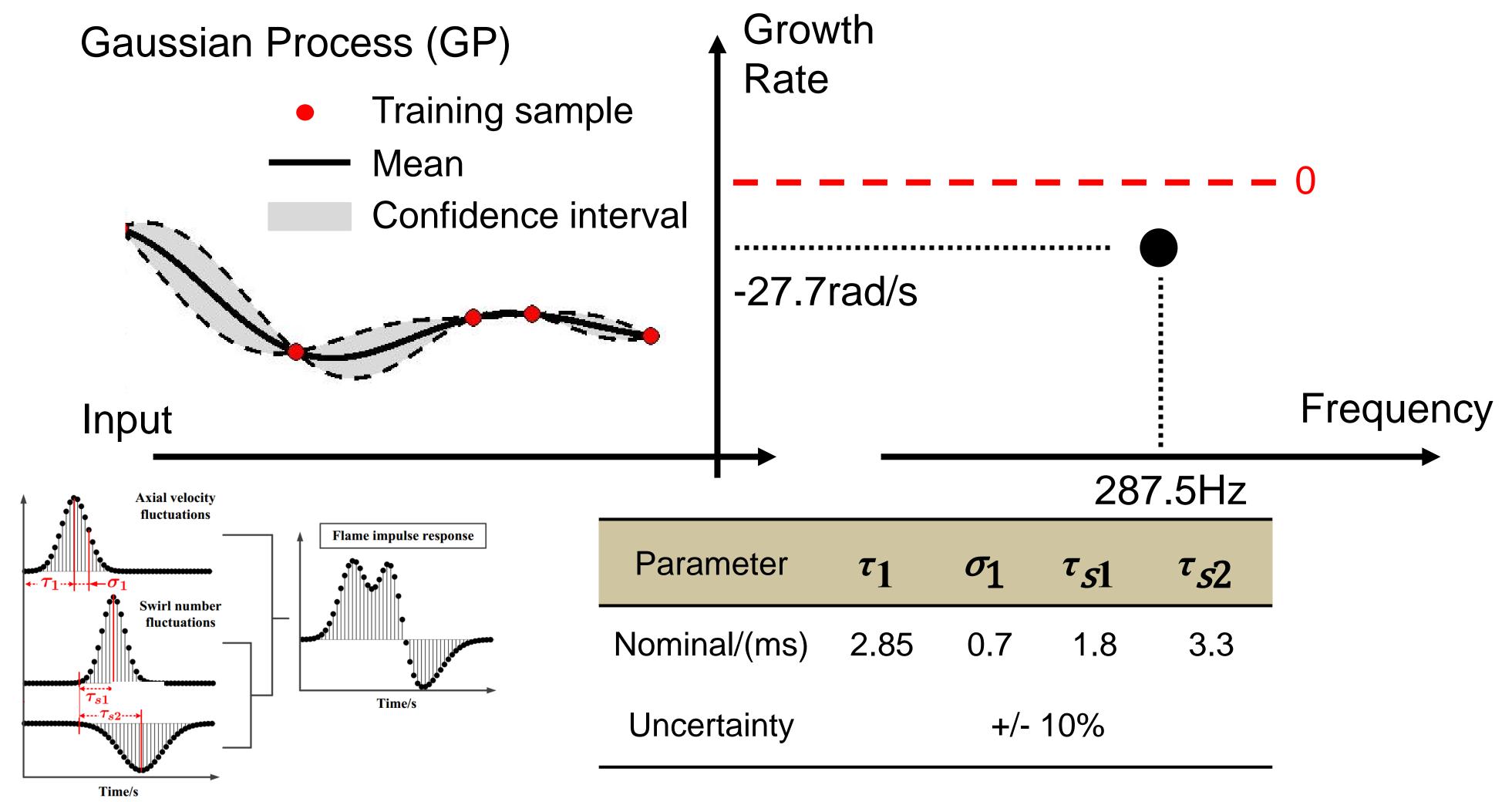
Network model, flame impulse response and Gaussian process



[1] Komarek, T., Polifke, W., 2010, J Eng Gas Turbines Power.



Network model, flame impulse response and Gaussian process



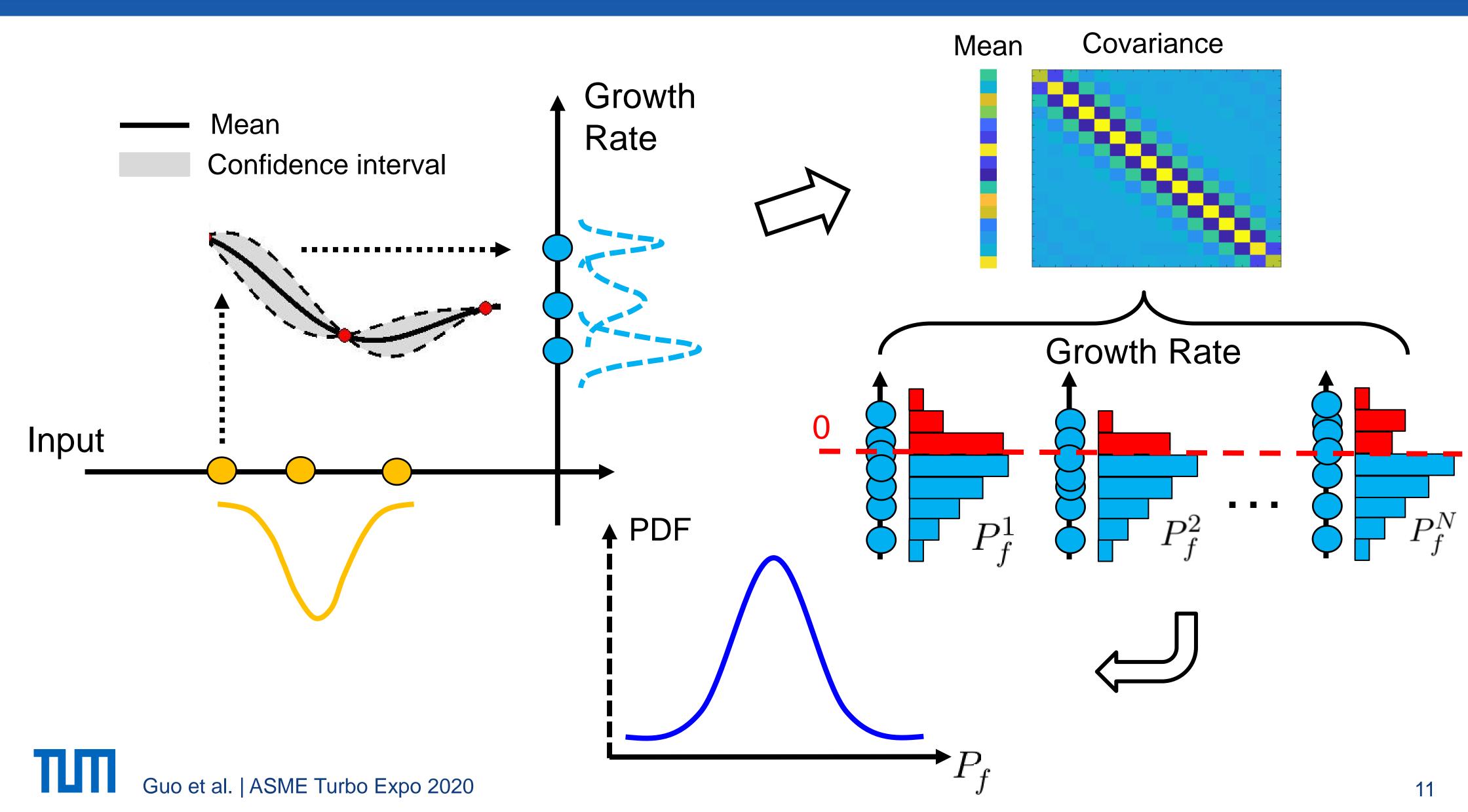
[1] Komarek, T., Polifke, W., 2010, J Eng Gas Turbines Power.



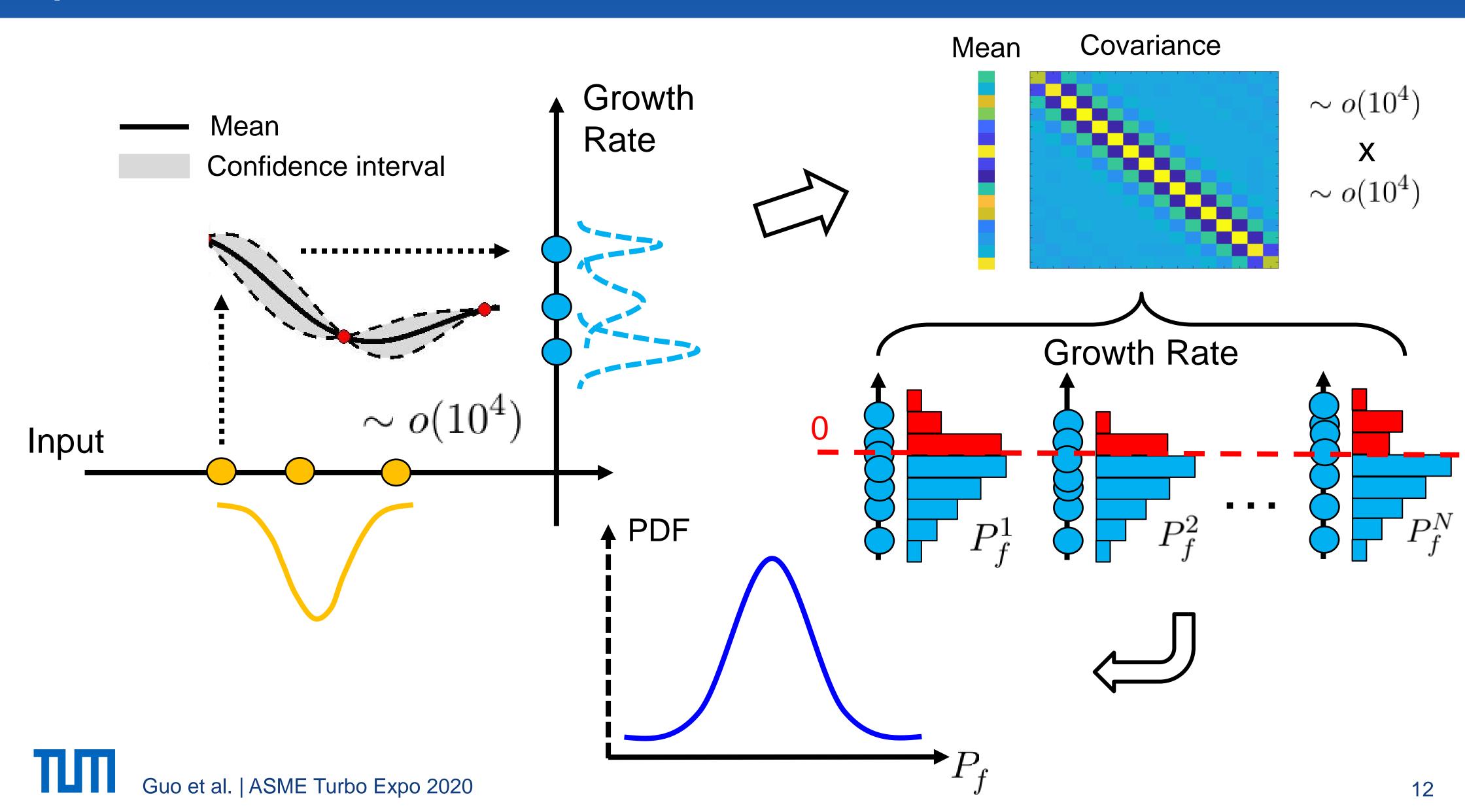
Presentation overview

- Motivation
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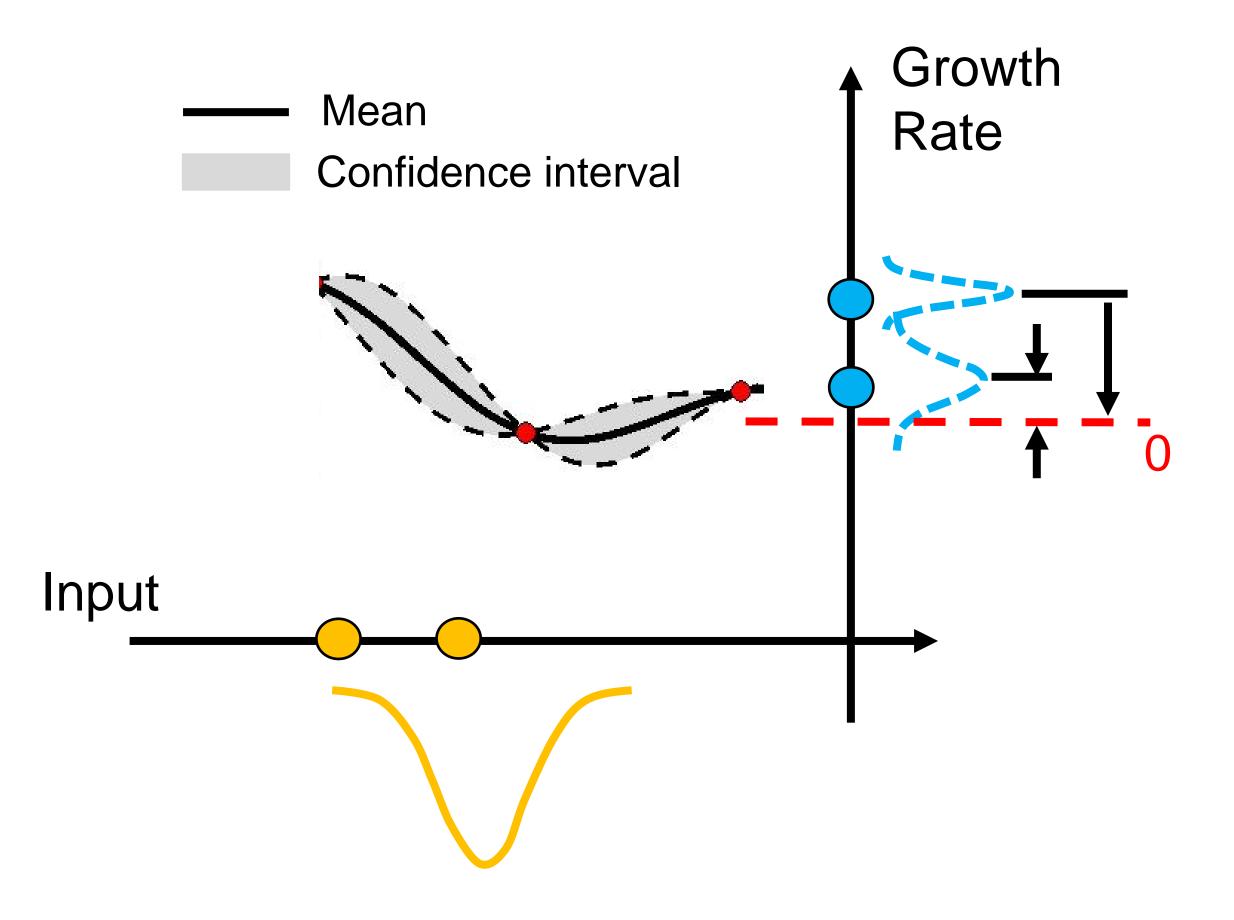
The growth rate prediction uncertainty yielded by GP permits the quantification of risk variations



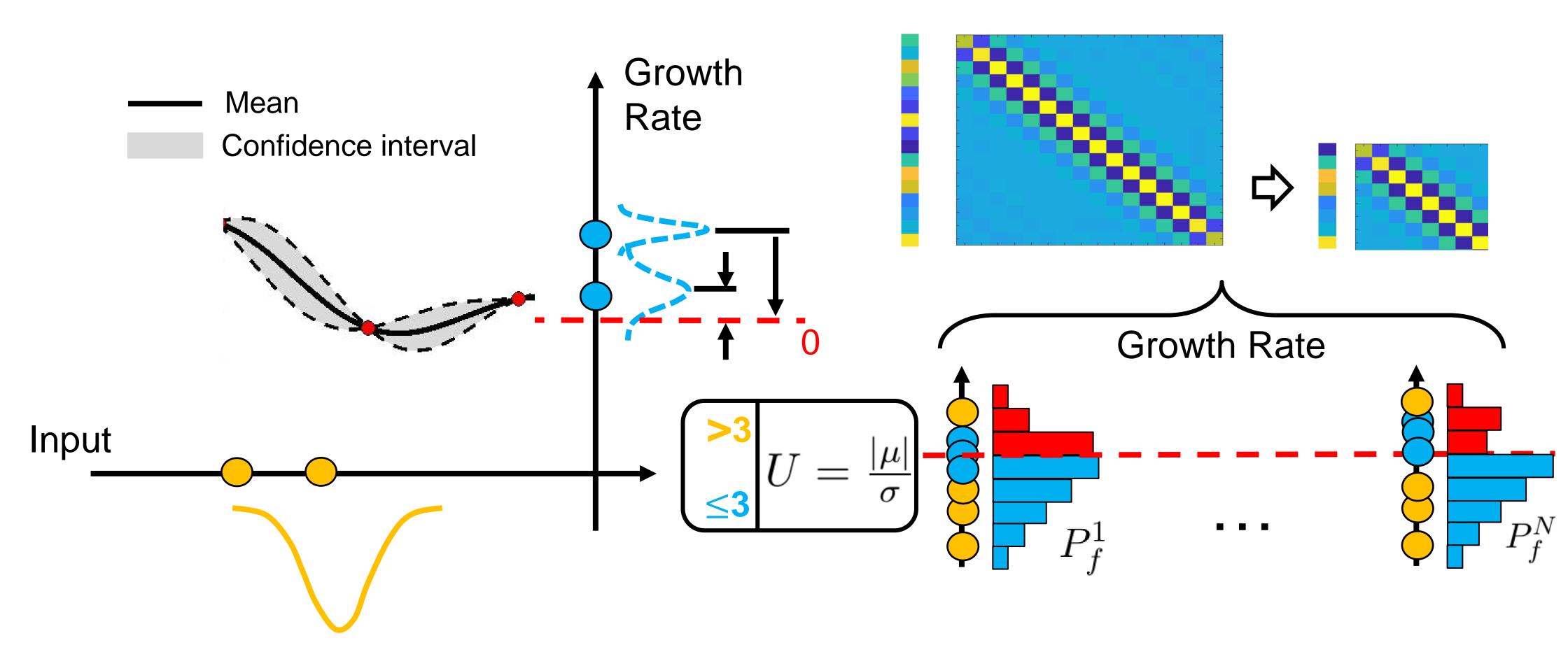
The growth rate prediction uncertainty yielded by GP permits the quantification of risk variations



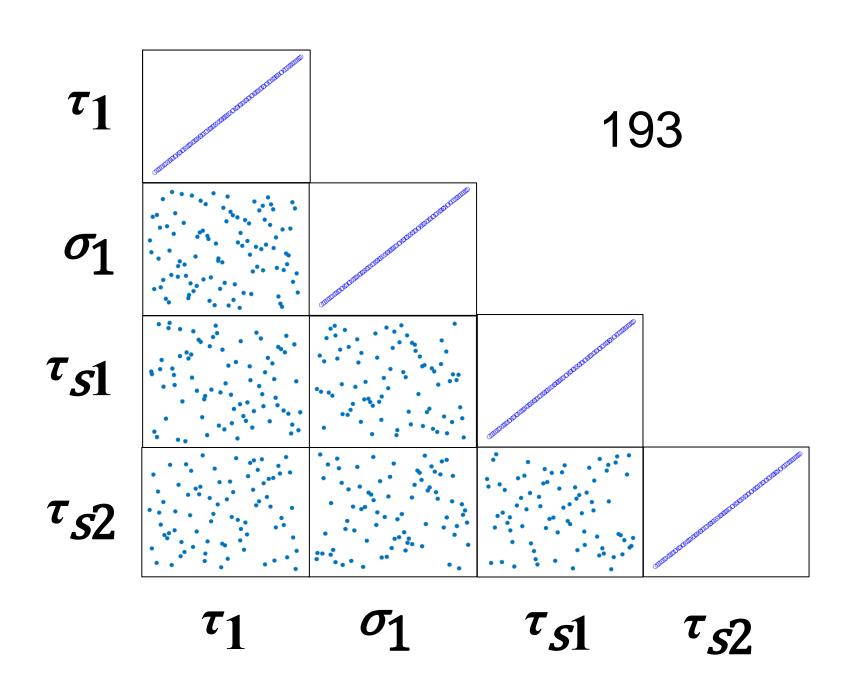
Only outputs close to stability margin worth considering in quantifying risk variations

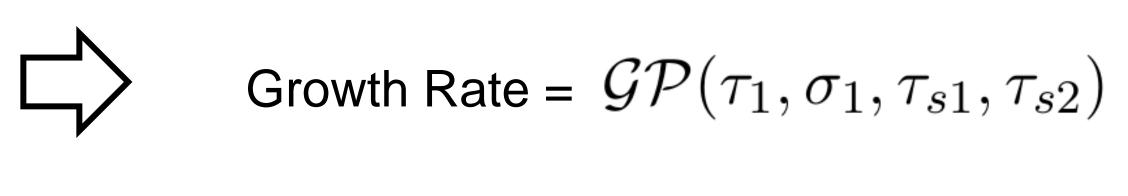


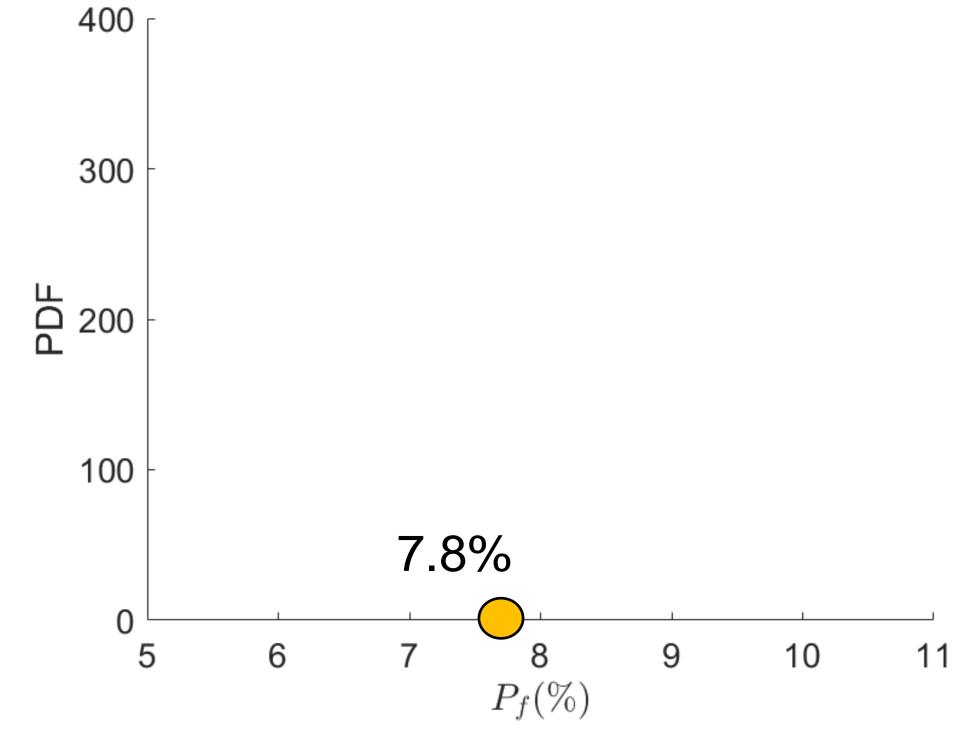
Only outputs close to stability margin worth considering in quantifying risk variations



Taking GP uncertainty into account yields a more robust risk calculation

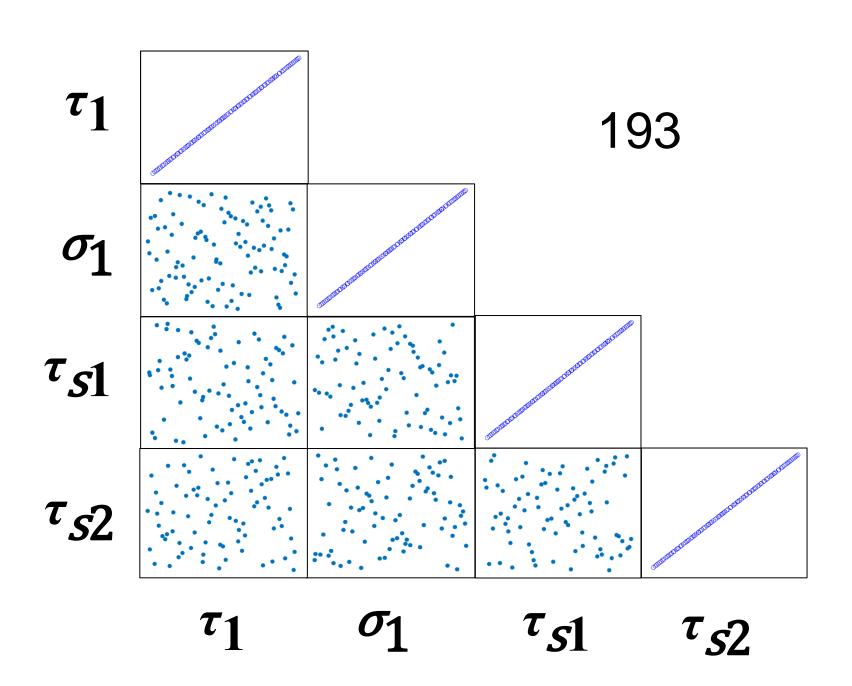


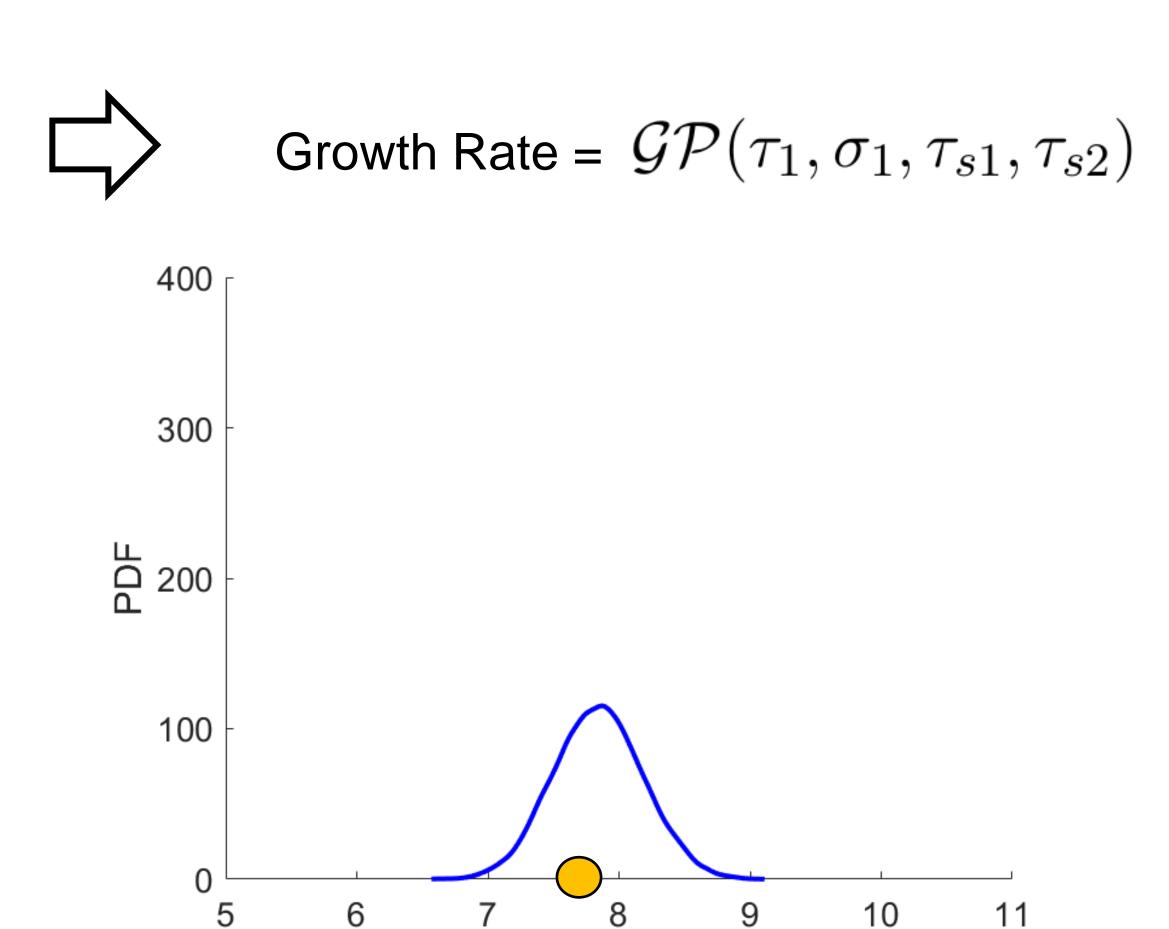






Taking GP uncertainty into account yields a more robust risk calculation

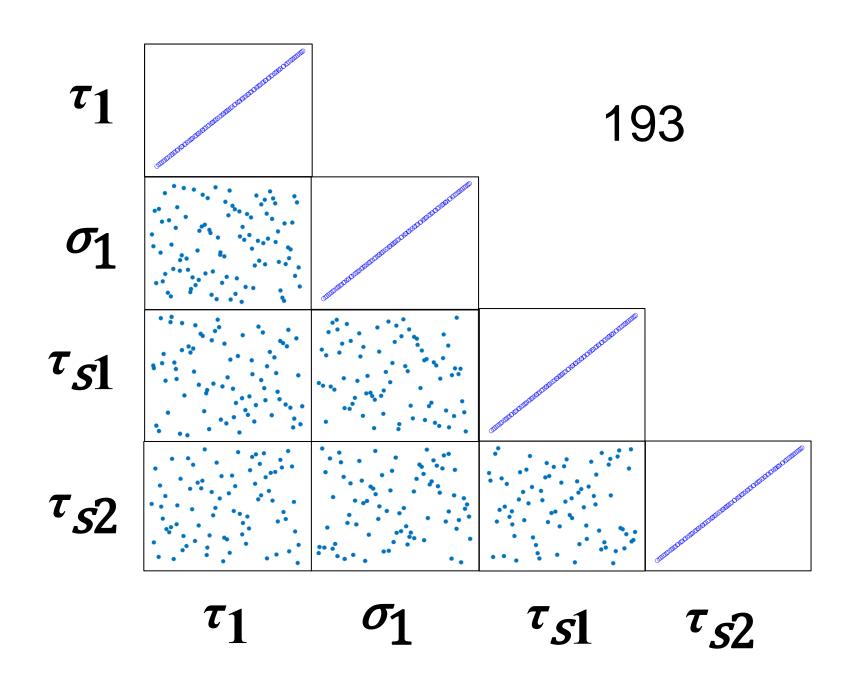


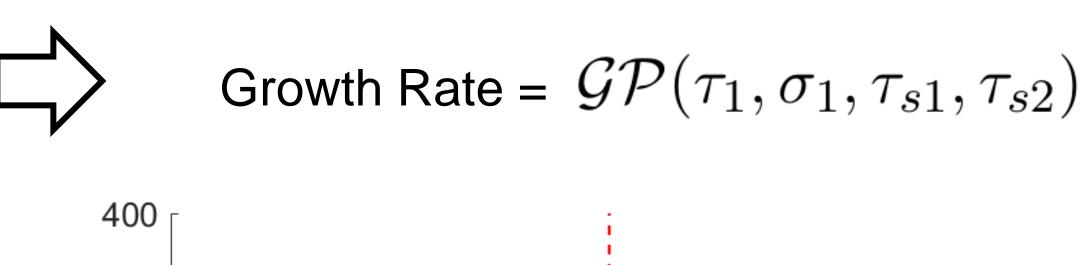


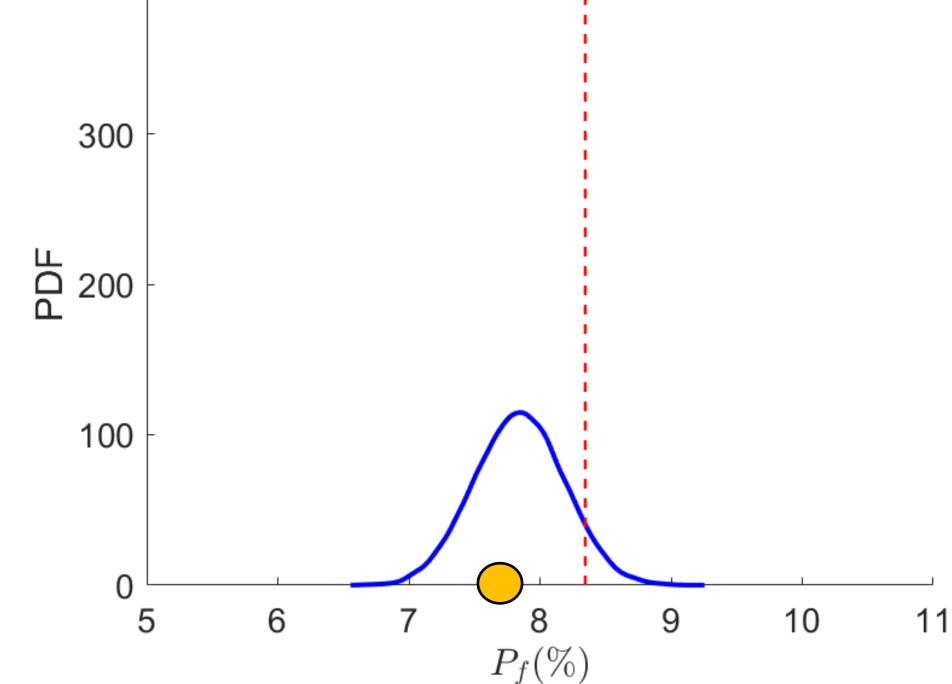
 $P_f(\%)$



Taking GP uncertainty into account yields a more robust risk calculation

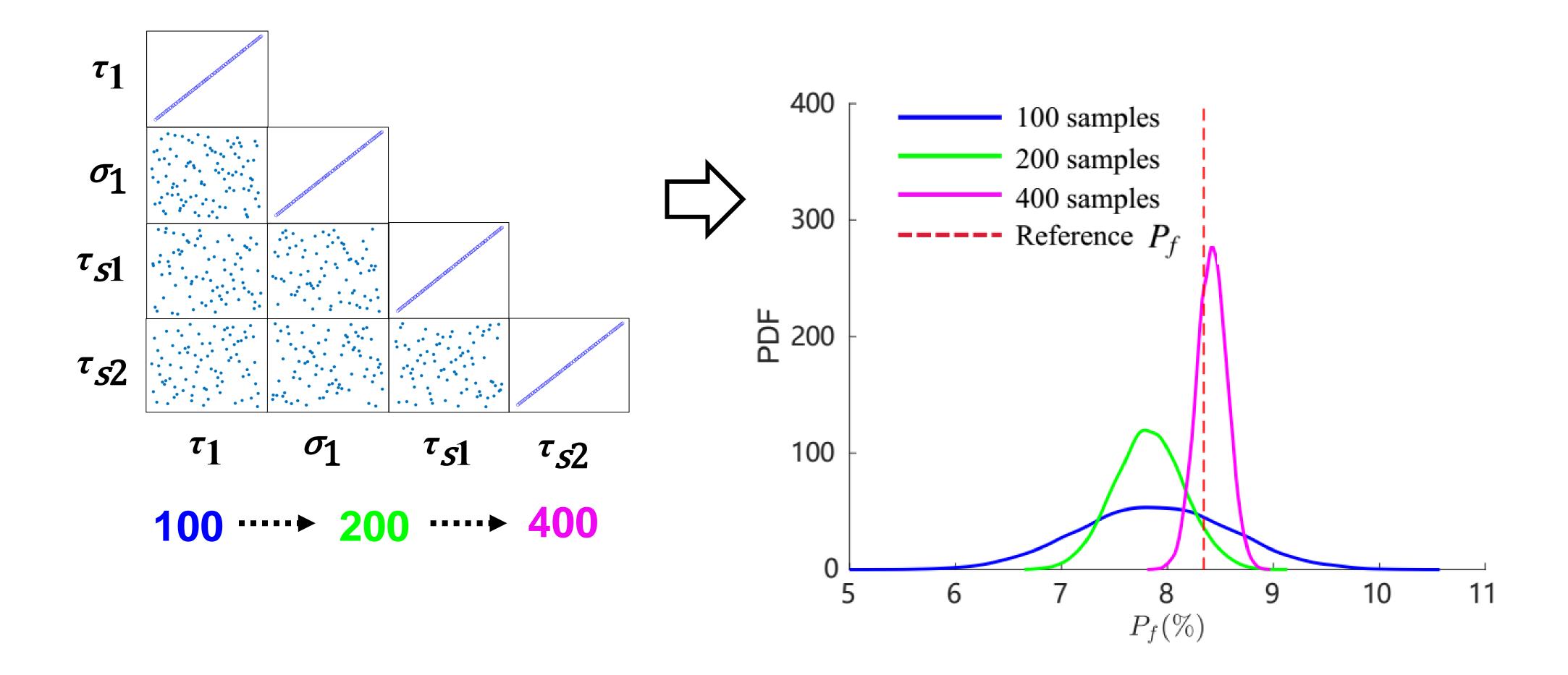






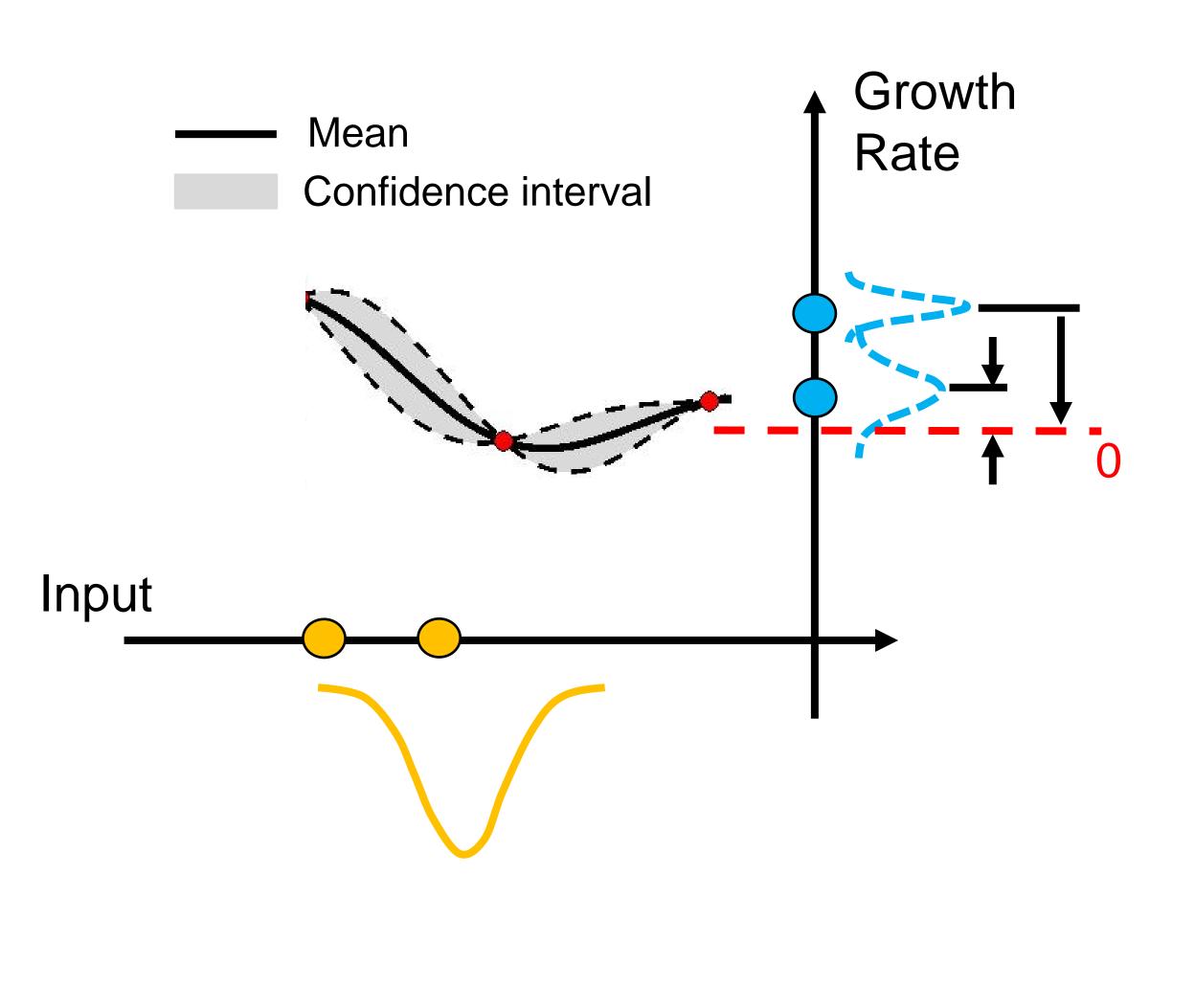


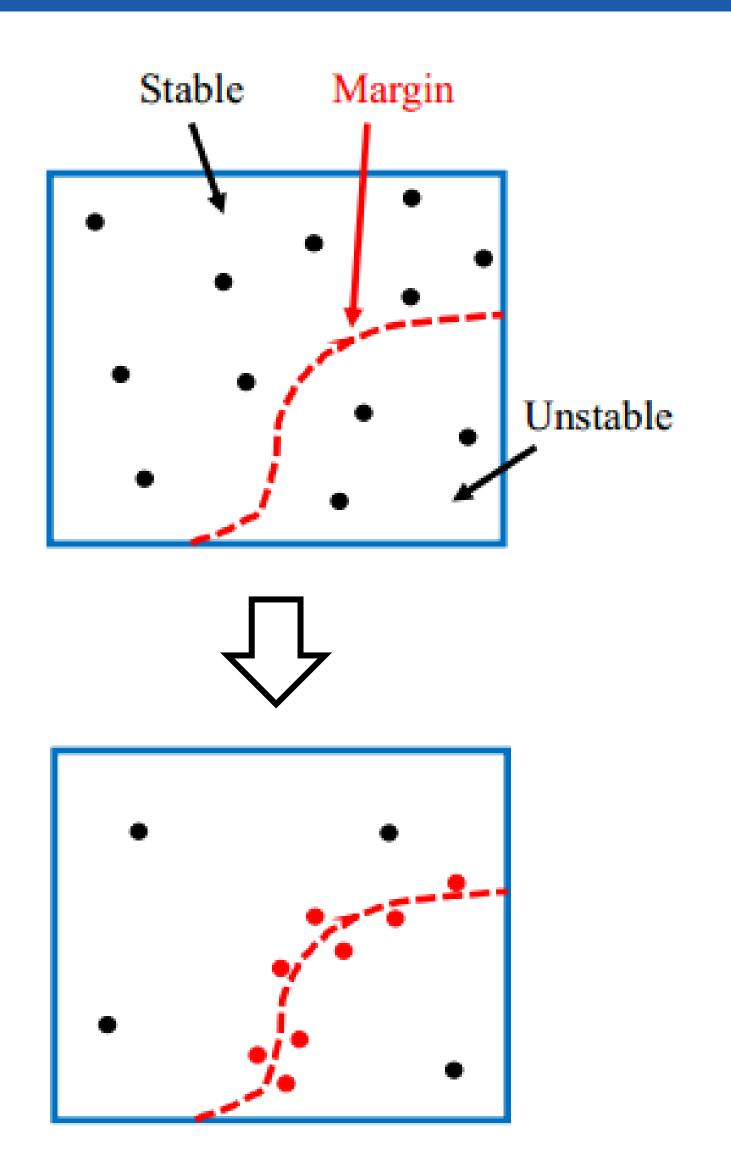
More training samples lead to more robust risk estimation





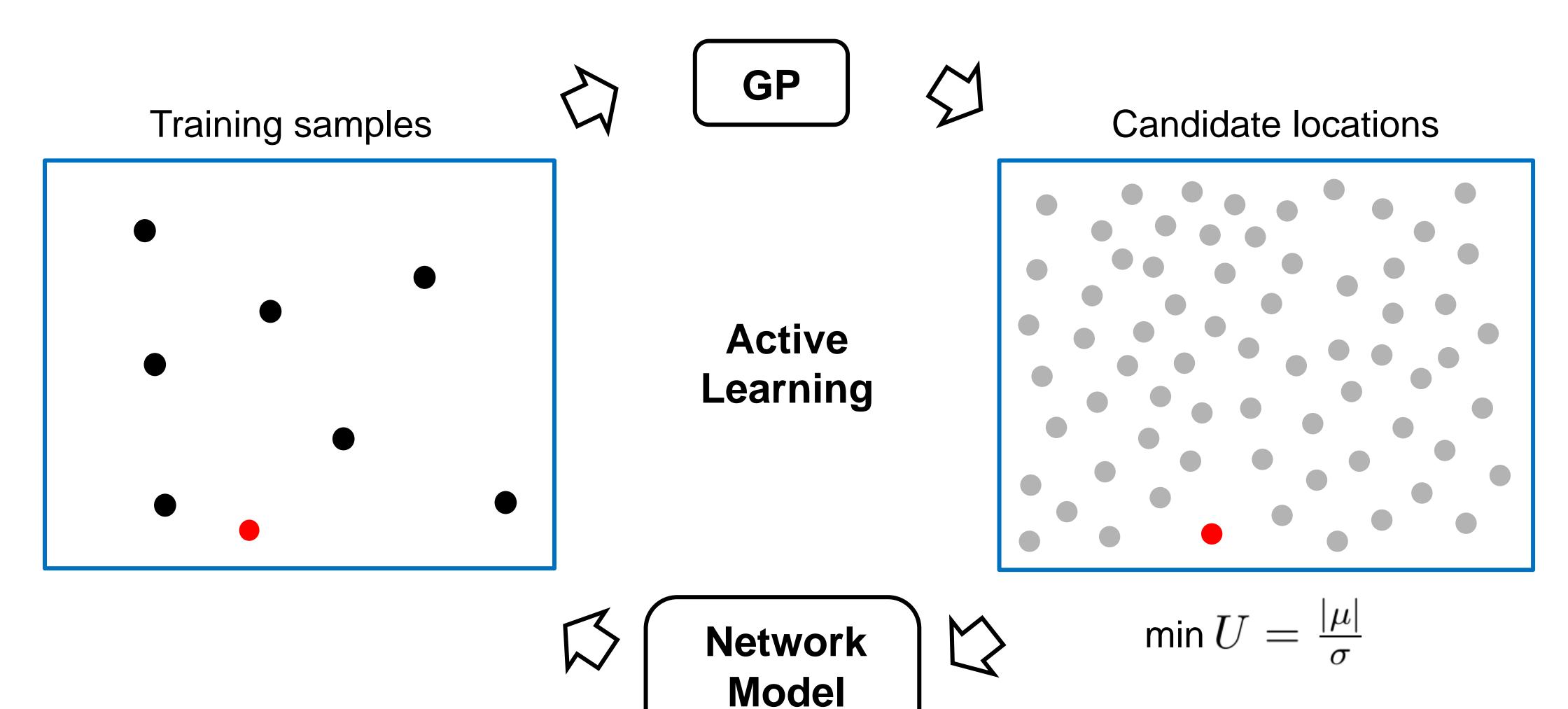
More training samples should be allocated in the vicinity of the stability margin





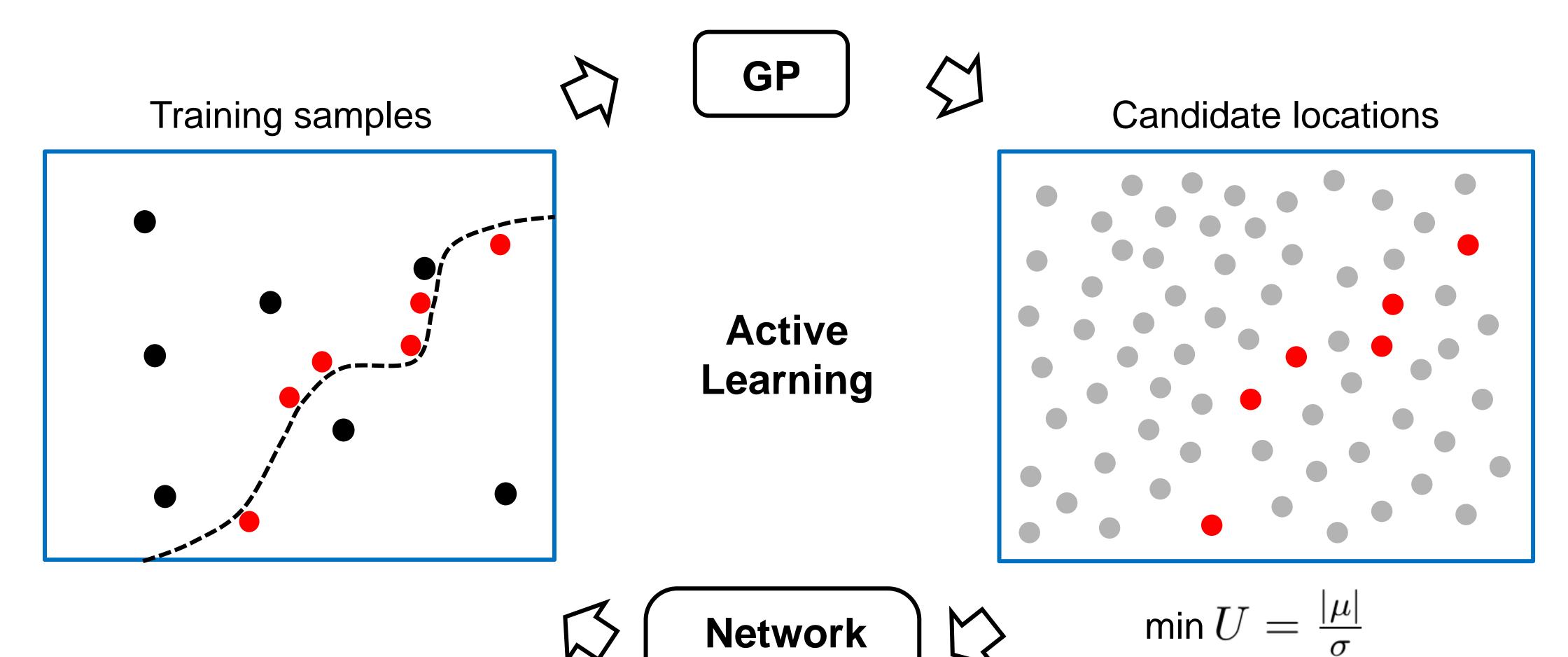


An active learning scheme is adopted to enrich the training dataset



Calculation

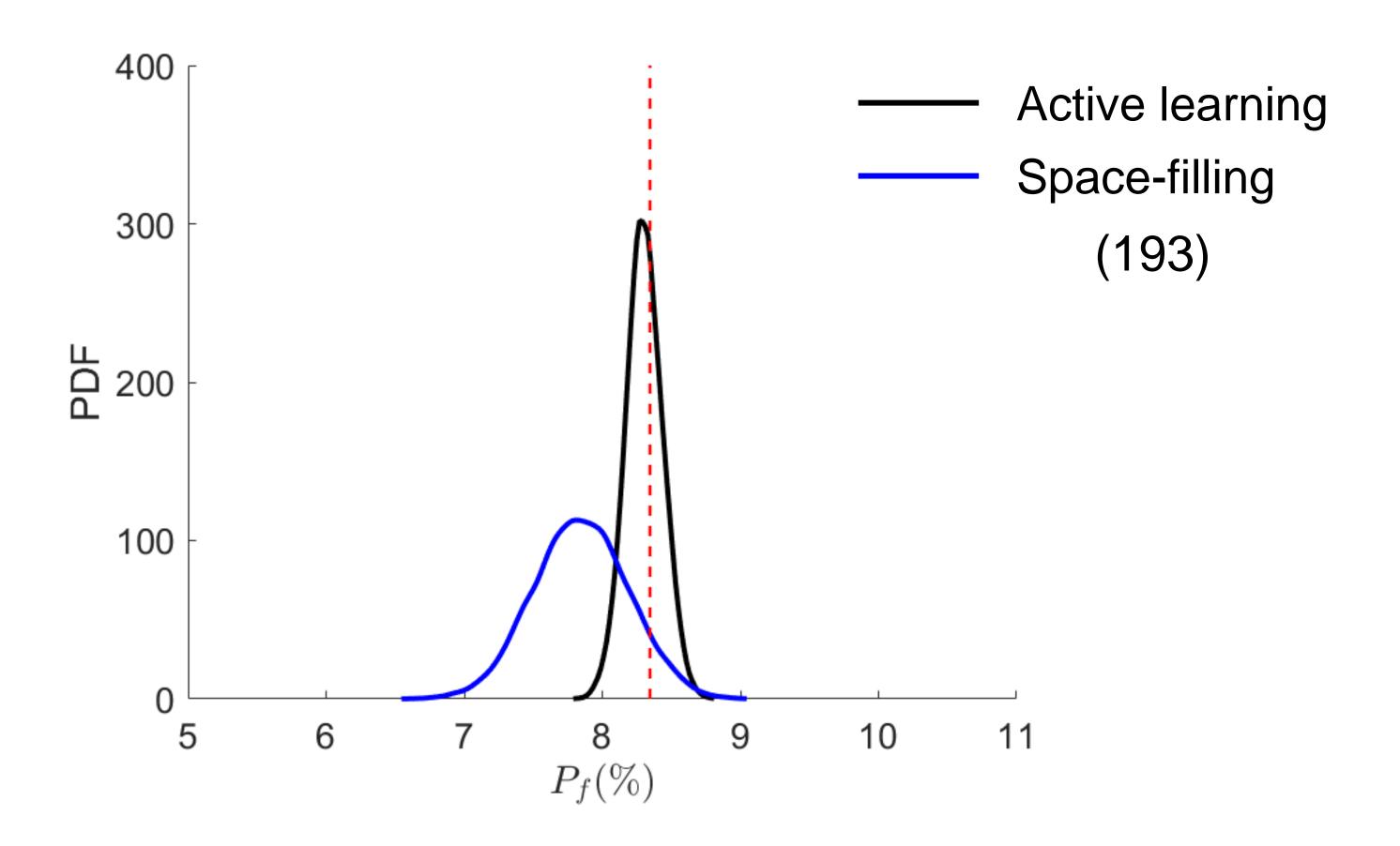
An active learning scheme is adopted to enrich the training dataset



Model

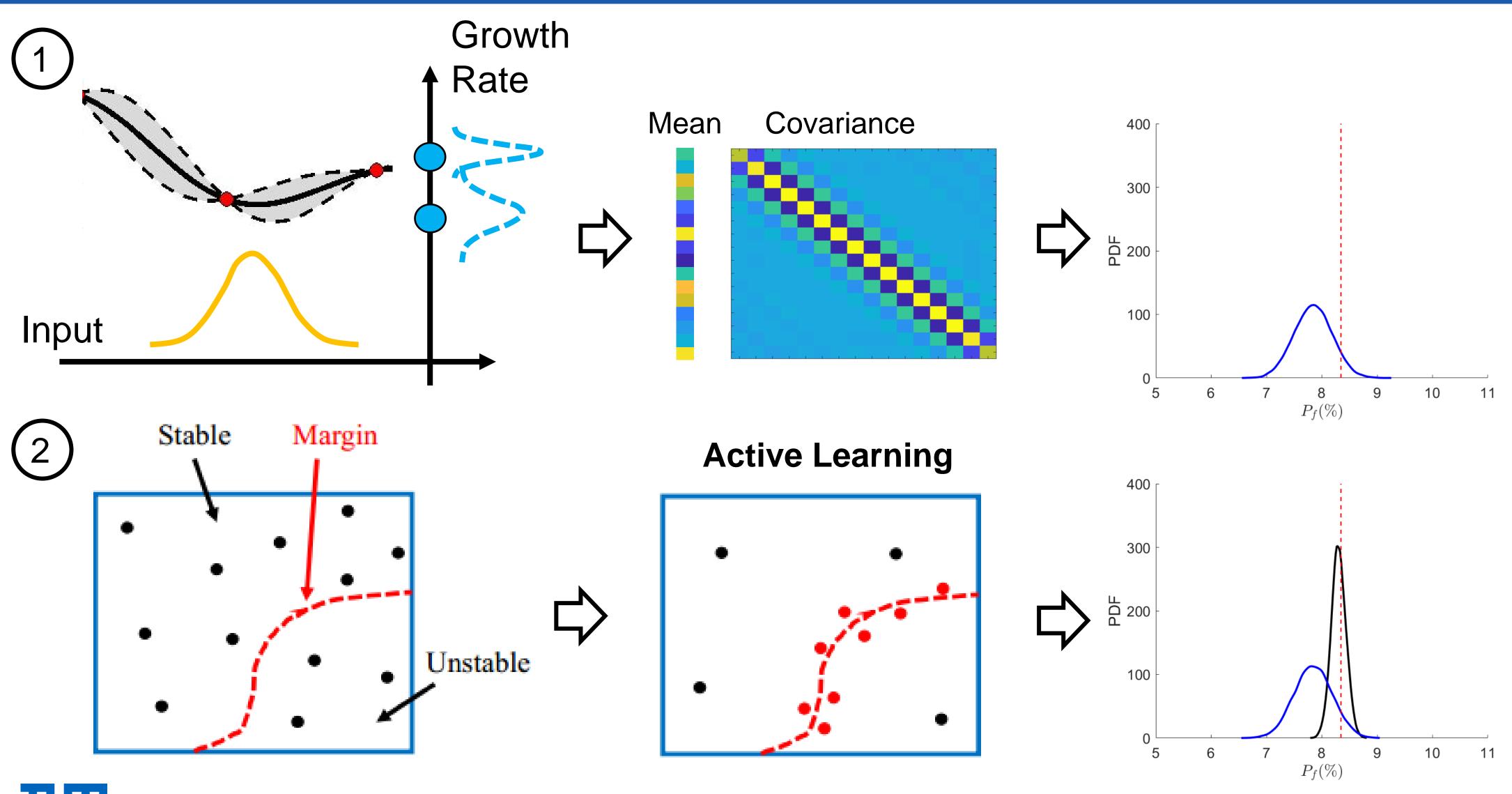
Calculation

Active learning scheme delivers a more robust risk estimation given the same number of training samples





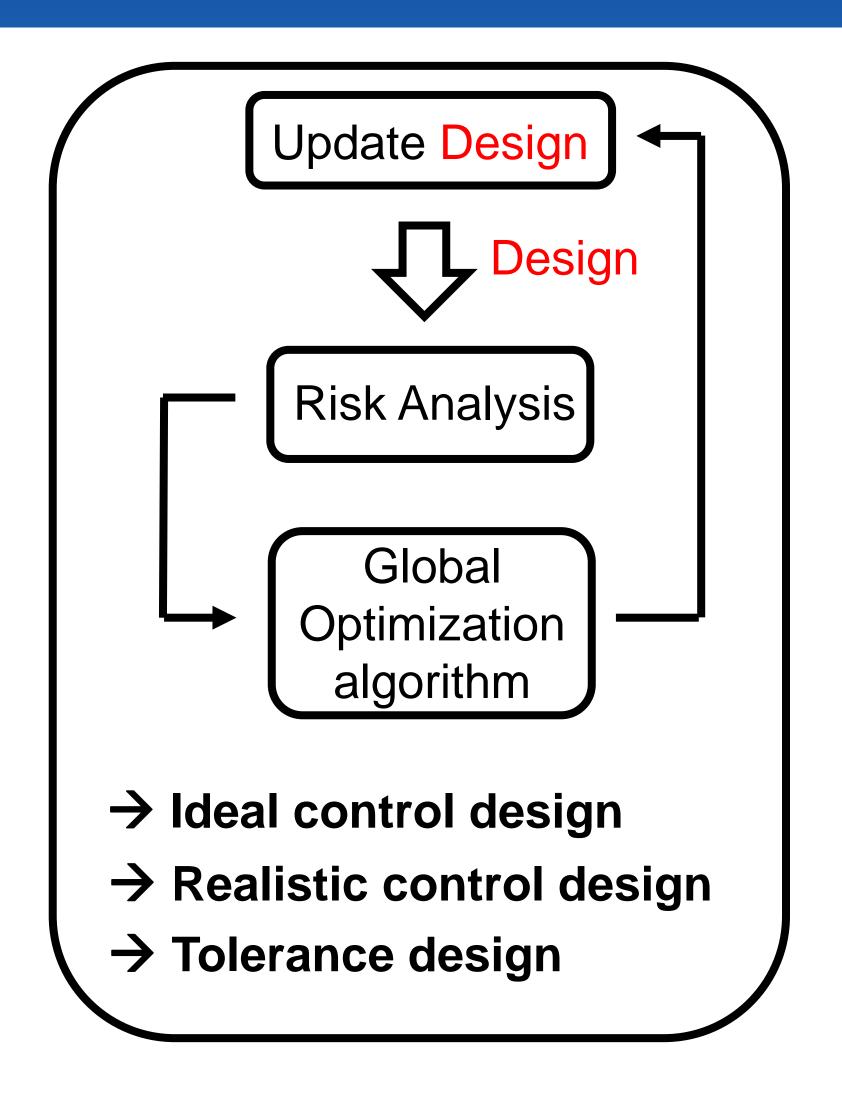
Conclusion



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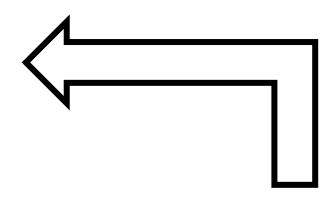


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Reliable calculation of thermoacoustic instability risk using an imperfect surrogate model

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Quantify risk calculation variation

Reduce risk calculation variation



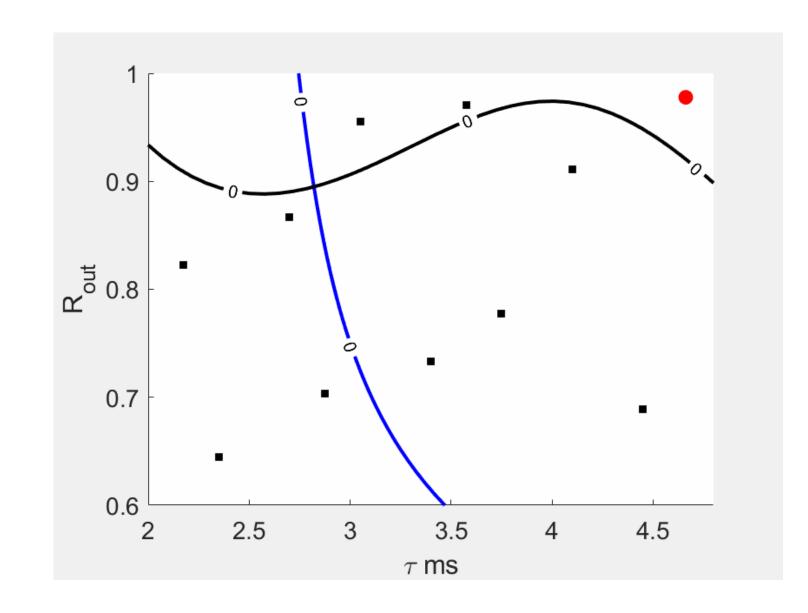
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Calculating thermoacoustic system instability risk using imperfect surrogate models



Improved training scheme



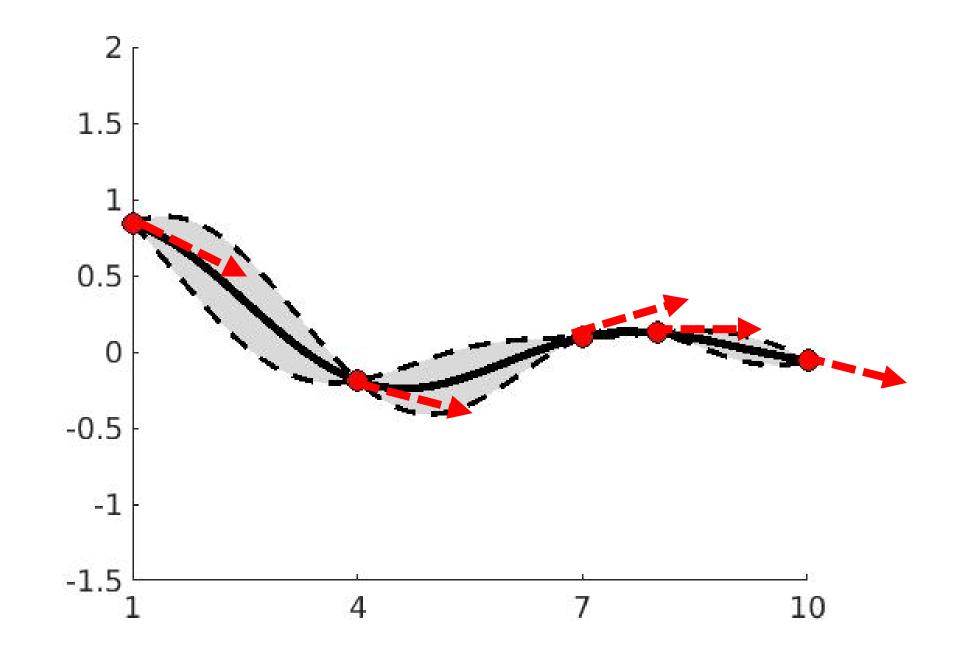
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Reliable calculation of thermoacoustic instability risk using an imperfect surrogate model

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Calculating thermoacoustic system instability risk using imperfect surrogate models



- Improved training scheme
- Gradient-enhanced GP

Efficient robust design for thermoacoustic instability analysis: A Gaussian Process approach

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Reliable calculation of thermoacoustic instability risk using an imperfect surrogate model

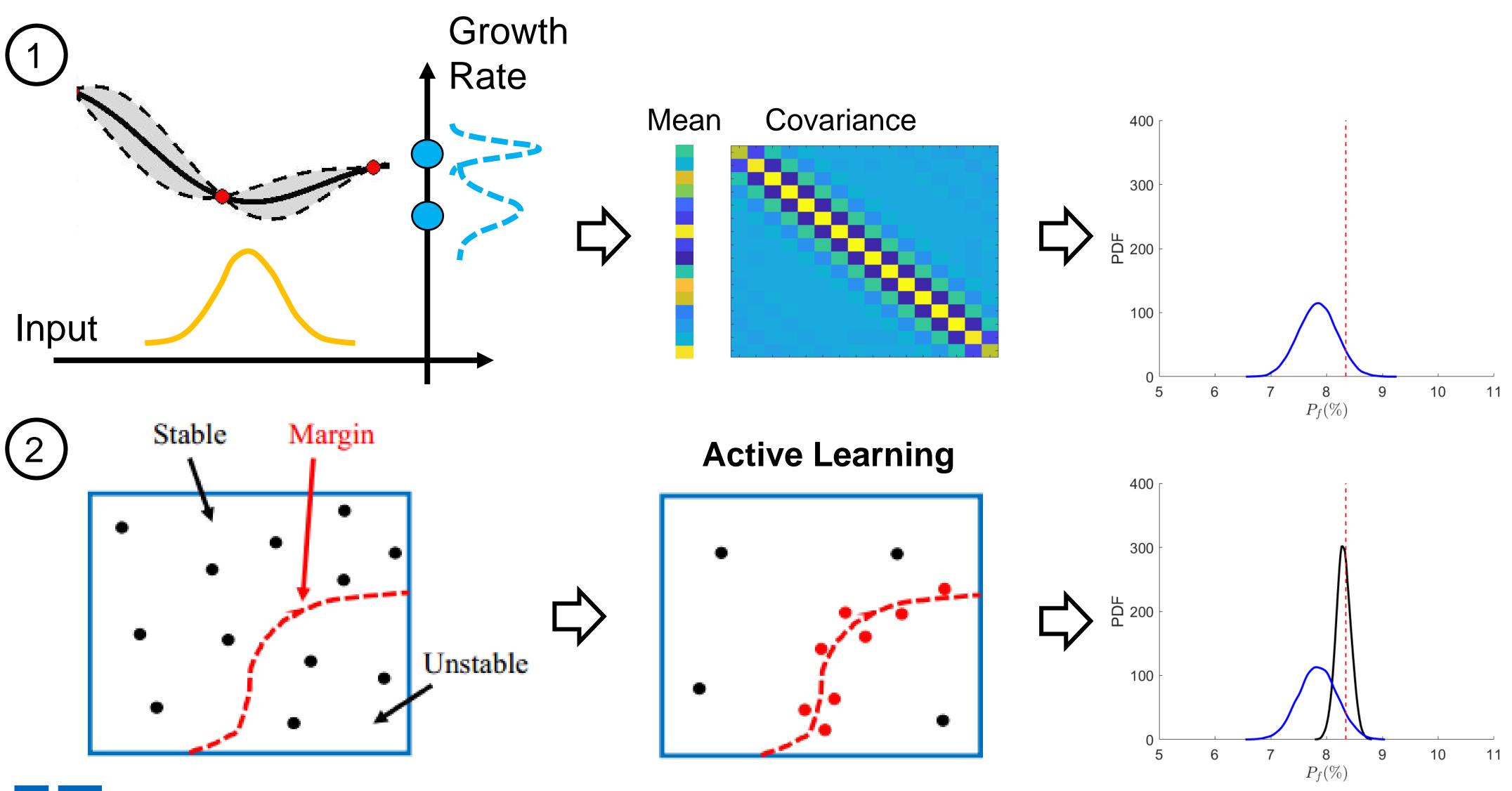
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Calculating thermoacoustic system instability risk using imperfect surrogate models

- Improved training scheme
- Gradient-enhanced GP
- Helmholtz solver instead of network model
- More uncertain parameters and larger variational ranges



Conclusion



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