

Ruprecht-Karls-Universität Heidelberg

# **Pre-talk : Bayesian asymptotics and nonparametric inverse problems**

*In anticipation of Bartek Knapik talk  
Xavier Loizeau*

30<sup>th</sup> of June 2016



Frequentist paradigm : convergence rates

Bayesian formulation : concentration rate

Inverse problems



Frequentist paradigm : convergence rates

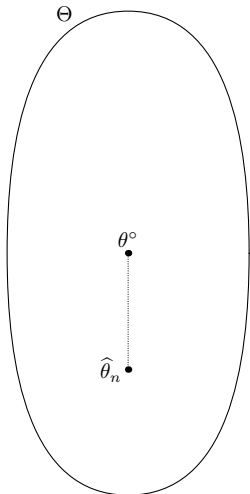
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# Point-wise comparison of frequentist estimators



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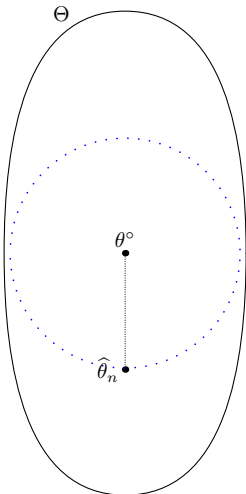


# Point-wise comparison of frequentist estimators

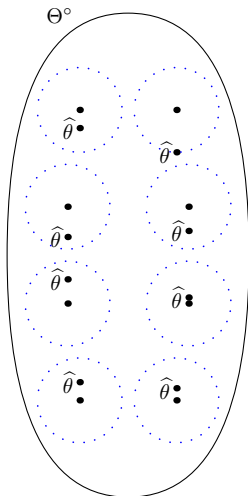


- Measure the performance of a frequentist estimator  $\hat{\theta}_n$  using **quadratic risk** for a given parameter  $\theta^\circ$

$$\mathbb{E}_{\theta^\circ} \left[ d(\hat{\theta}_n, \theta^\circ)^2 \right]$$



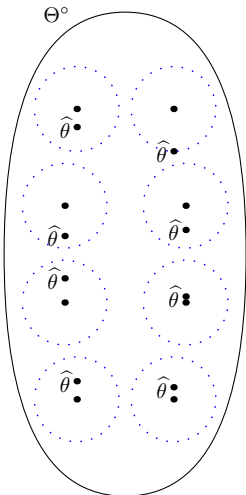
# Comparison of frequentist estimators over a subset



# Comparison of frequentist estimators over a subset



- Measure the performance of a frequentist estimator using **maximal risk** over a class  $\Theta^\circ$  of parameters

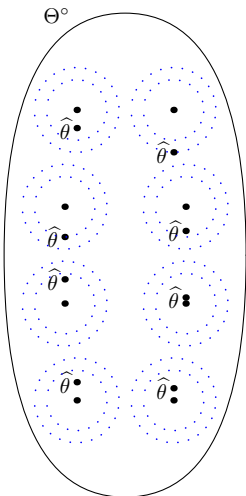


$$\sup_{\theta^\circ \in \Theta^\circ} \mathbb{E}_{\theta^\circ} \left[ d(\hat{\theta}, \theta^\circ)^2 \right]$$

# Comparison of frequentist estimators over a subset



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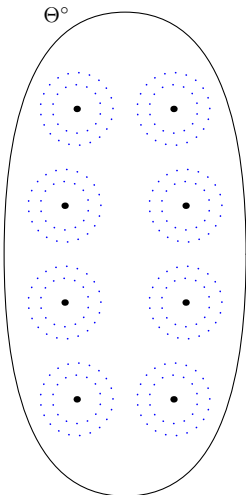
$$\sup_{\theta^\circ \in \Theta^\circ} \mathbb{E}_{\theta^\circ} \left[ d(\hat{\theta}, \theta^\circ)^2 \right]$$

- Goal : finding a lower bound  $\mathcal{R}_n^*(\Theta^\circ) \dots$

$$\inf_{\tilde{\theta}} \sup_{\theta^\circ \in \Theta^\circ} \mathbb{E}_{\theta^\circ} \left[ d(\tilde{\theta}, \theta^\circ)^2 \right] \geq C_1 \cdot \mathcal{R}_n^*(\Theta^\circ)$$



# Comparison of frequentist estimators over a subset



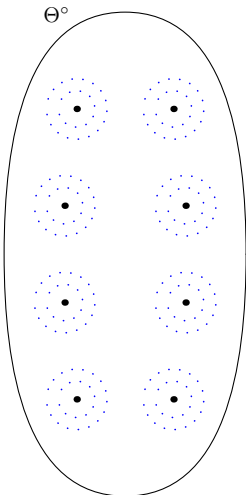
- Measure the performance of a frequentist estimator using **maximal risk** over a class  $\Theta^\circ$  of parameters

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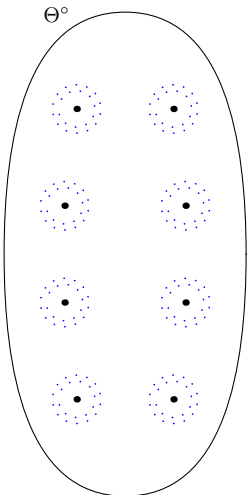
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# Comparison of frequentist estimators over a subset



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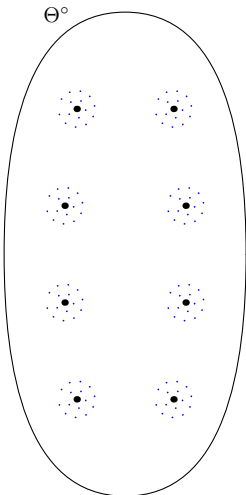
- Goal : finding a lower bound  $\mathcal{R}_n^\star(\Theta^\circ) \dots$

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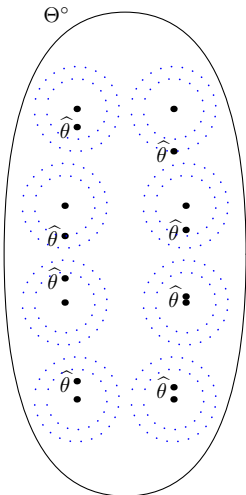
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- ... which is reached by an estimator  $\hat{\theta}$

$$\sup_{\theta^\circ \in \Theta^\circ} \mathbb{E}_{\theta^\circ} \left[ d(\hat{\theta}, \theta^\circ)^2 \right] \leq C_2 \cdot \mathcal{R}_n^*(\Theta^\circ)$$



Frequentist paradigm : convergence rates

Bayesian formulation : concentration rate

Inverse problems



Let  $(\Theta, \mathfrak{A})$  be a measurable space

$\theta$  is a random variable  $(\Omega, \mathfrak{F}) \rightarrow (\Theta, \mathfrak{A})$

$$\theta \sim \Pi$$

Denote by  $p_\theta$  the density of  $\Pi$  with respect to a measure  $\mu$

Posterior distribution

$$\forall \mathfrak{B} \in \mathfrak{A} \quad \Pi_{\theta|Y}(\mathfrak{B}) = \frac{\int_{\mathfrak{B}} p_\theta(Y) d\Pi(\theta)}{\int_{\Theta} p_\theta(X) d\Pi(\theta)}$$

# Bayesian paradigm

Comparing prior choices



Taking a frequentist point of view

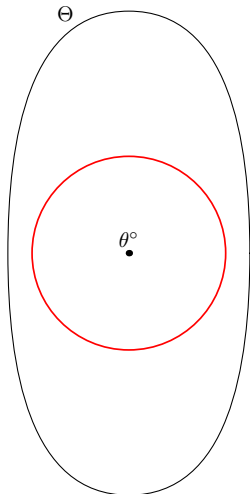
- ▶  $\theta^\circ$  the true parameter
- ▶ Is  $\Pi_{\theta|Y}$  shrinking around  $\theta^\circ$  as  $\epsilon$  tends to 0?
- ▶ How fast?





- Concentration rate  $(\phi_\epsilon)_{\epsilon \in \mathbb{R}_+}$

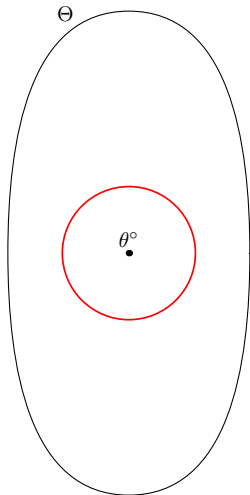
$$\exists c \in \mathbb{R}_+, \quad \lim_{\epsilon \rightarrow 0} \mathbb{E}_{\theta^\circ} \left[ \Pi_{\theta|Y} \left( d(\theta, \theta^\circ)^2 \geq c\phi_\epsilon \right) \right] = 0$$





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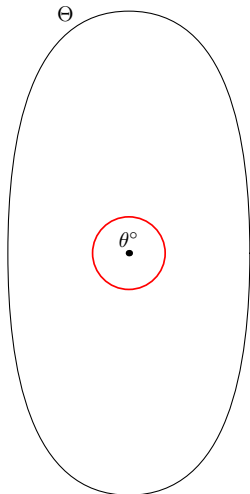
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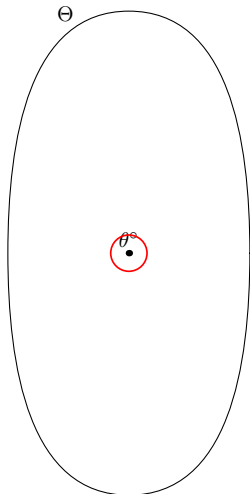
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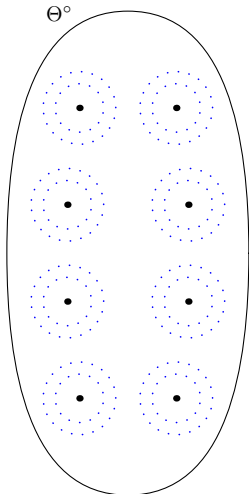




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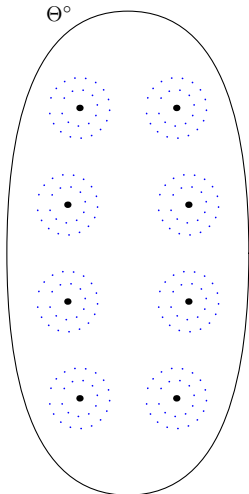


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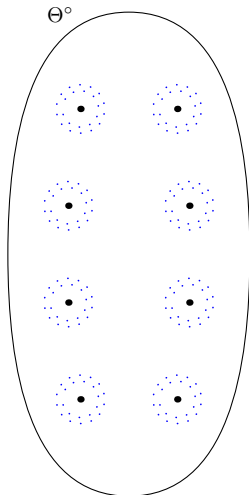


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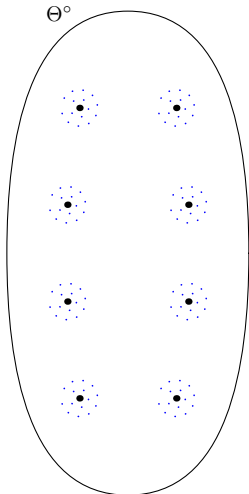


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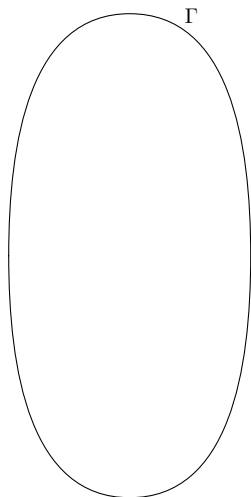
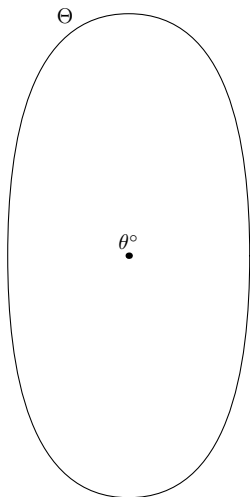




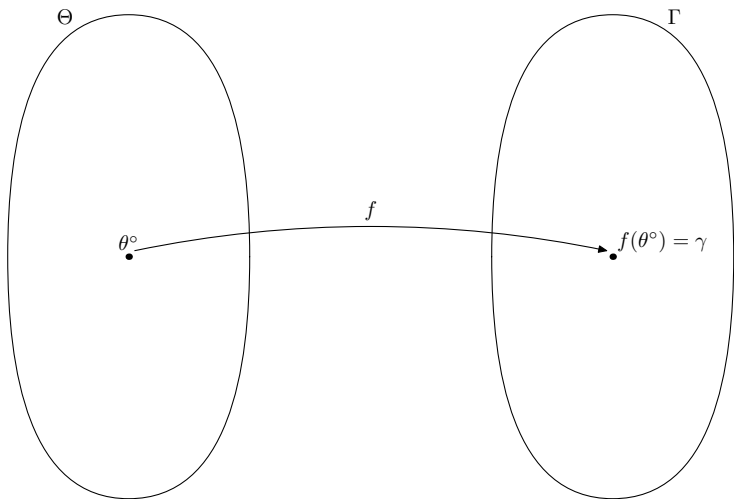
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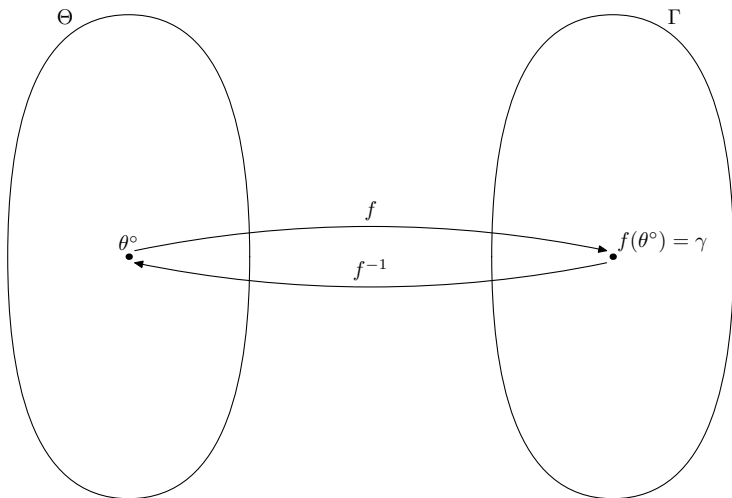
Inverse problems



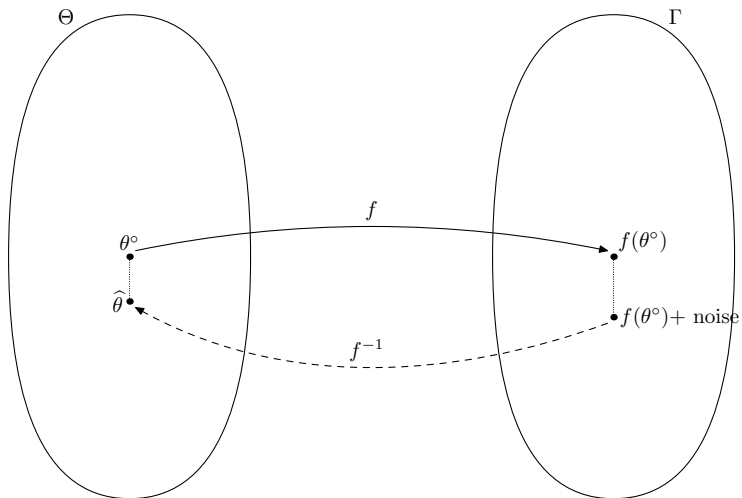
# Inverse problems



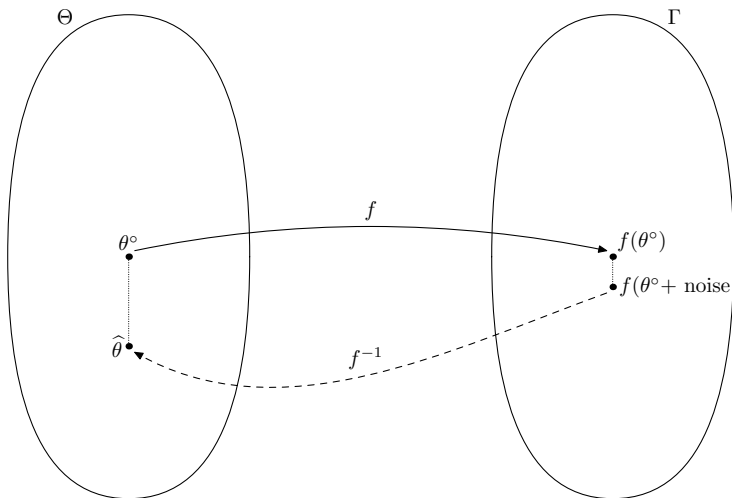
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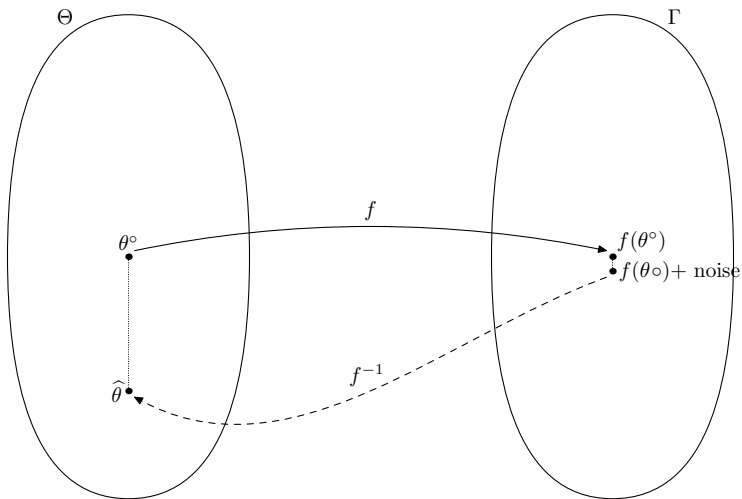
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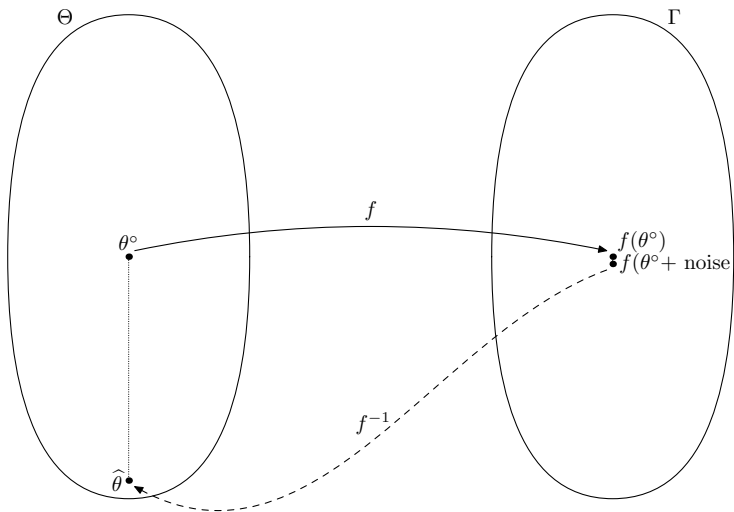
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