kernelPSI: a Post-Selection Inference Framework for Nonlinear Variable Selection

ICML 2019, Long Beach

joint work with

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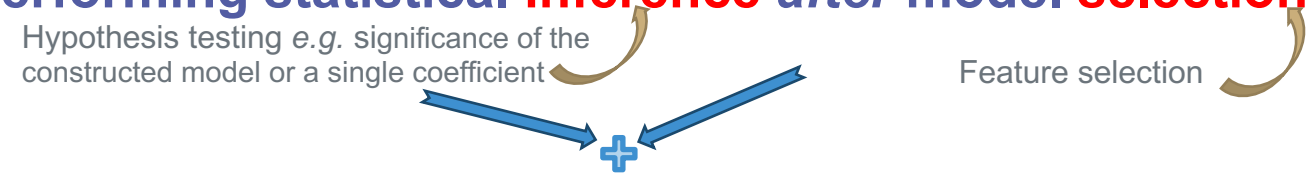
Post-Selection Inference (PSI)

Why does it matter?

- **PSI: performing statistical inference *after* model selection**



Need to *account* for the selection event for *valid* inference!

In *practice*,

- Data splitting: loss of accuracy in selection and power in inference

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- Exact PSI: new momentum thanks to the *polyhedral lemma*, several setups covered

Absence of nonlinear effects and interactions among covariates!

kernelPSI: key ideas

A generalization of linear methods

Select a subset of kernels that are most associated with an outcome Y, and then to **measure** the significance of their association with Y (individually or in a joint manner)

Quadratic kernel association scores

$$s(\mathrm{K},\mathrm{Y}) = \mathrm{Y}^{\mathrm{T}}\mathrm{Q}(\mathrm{K})\mathrm{Y}$$

- Prototypes: for Q p.d., $s(\mathrm{K},\mathrm{Y}) = \left\|\widehat{\mathrm{Y}}_{\mathrm{K}}\right\|^2$: kernel PCA and kernel ridge regression
- $s(K,Y) = \widehat{HSIC}\ (K, YY^T)$

For a quadratic kernel association score, select a subset of kernels by:

- Filtering
- Forward/backward stepwise selection

Selection

Selection event = a conjunction of quadratic constraints

In a model of the form $Y = \mu + \sigma^2\epsilon$, test for a null hypothesis:

$$H_0: s(K,\mu) = 0$$

- Test statistics: norm of prototypes, log-likelihoods
- Constrained sampling to generate empirical p-values

Inference

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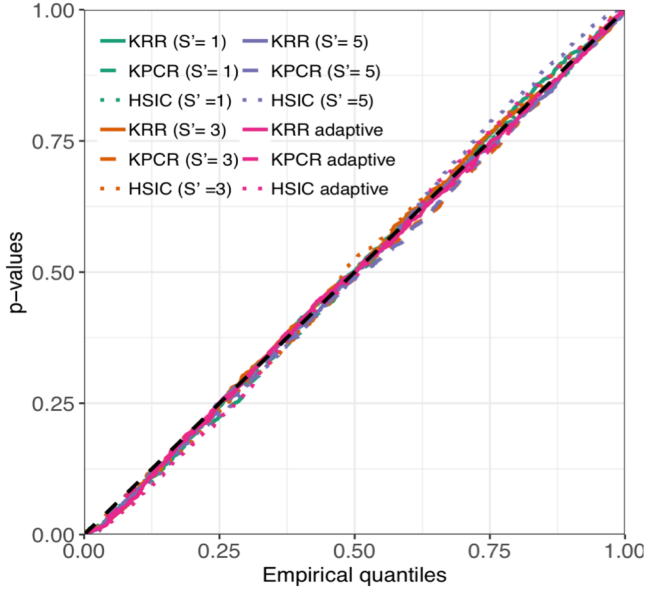
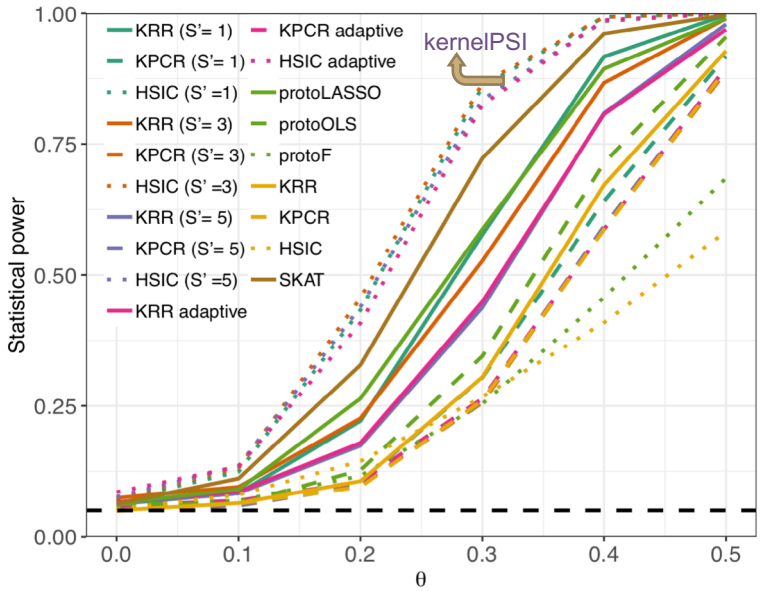
kernelPSI: results





The obtained p-values are valid!

KernelPSI is better than non-selective kernel baselines and selective linear ones!

kernelPSI: poster #228

THANK YOU!

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