

# Recap

## High-dimensional additive models

$$Y_i = \mu + \sum_{j=1}^p f_j^0(X_i^{(j)}) + \varepsilon_i \quad (i = 1, \dots, n; \quad p \gg n)$$

$$f_j^0 : \mathbb{R} \rightarrow \mathbb{R} \text{ smooth, } \mathbb{E}[f_j^0(X_i^{(j)})] \equiv 0 \quad \forall j$$

aim: estimator such that either  $\hat{f}_j(.) \equiv 0$  or  $\hat{f}_j(.)$  is not the zero function

$\leadsto$  Group Lasso type problem

but: sparsity and **smoothness** penalty (SpS) is better than Group-sparsity (with respect to basis expansion) alone

## Sparsity Smoothness (SpS) penalty with smoothing spline

$$\hat{f}_1, \dots, \hat{f}_p = \operatorname{argmin}_{f_1, \dots, f_p \in \mathcal{F}} \left( \|Y - \sum_{j=1}^p f_j\|_n^2 + \lambda_1 \sum_{j=1}^p \|f_j\|_n + \lambda_2 \sum_{j=1}^p I(f_j) \right)$$

$$\|f_j\|_n^2 = n^{-1} \sum_{i=1}^n |f_j(X_i^{(j)})|^2, \quad I(f_j)^2 = \int f_j''(x)^2 dx$$

where  $\mathcal{F}$  = Sobolev space of functions that are continuously differentiable with square integrable second derivatives

$$\hat{f}_1, \dots, \hat{f}_p = \operatorname{argmin}_{f_1, \dots, f_p \in \mathcal{F}} \left( \|Y - \sum_{j=1}^p f_j\|_n^2 + \lambda_1 \sum_{j=1}^p \|f_j\|_n + \lambda_2 \sum_{j=1}^p l(f_j) \right)$$

is a parametric problem of dimension  $d \approx 3pn$ , parametrized by natural cubic splines with basis functions encoded in a matrix

$$H_{n \times d} = (H_1, \dots, H_p), \quad (H_j)_{n \times \approx 3n} \text{ matrix}$$

and integrated squared second derivatives encoded in a matrix

$$(W_j)_{k,\ell} = \int h''_{j,k}(x) h''_{j,\ell}(x) dx$$

$\leadsto$

$$\hat{\beta} = \operatorname{argmin}_{\beta} \left( \|Y - H\beta\|_2^2/n + \lambda_1 \sum_{j=1}^p \sqrt{\beta_j^T H_j^T H_j \beta_j/n} + \lambda_2 \sum_{j=1}^p \sqrt{\beta_j^T W_j \beta_j} \right)$$

if the problem is sparse and smooth:

only a few  $X^{(j)}$ 's influence  $Y$  (only a few non-zero  $f_j^0$ ) and the non-zero  $f_j^0$  are smooth

$\leadsto$  one can often afford to model and fit additive functions in high dimensions

reason:

- ▶ dimensionality is of order  $\dim = O(pn)$   
 $\log(\dim)/n = O((\log(p) + \log(n))/n)$  which is still small
- ▶ sparsity **and** smoothness then lead to: if each  $f_j^0$  is twice continuously differentiable

$$\|\hat{f} - f^0\|_2^2/n = O_P(\underbrace{\text{sparsity}}_{\text{no. of non-zero } f_j^0} \sqrt{\log(p)} n^{-4/5})$$

(cf. Ch. 8.4 in Bühlmann & van de Geer (2011))