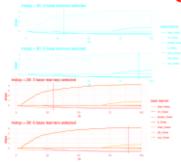
Introduction to Machine Learning

Boosting Boosting: CWB Basics 1

Gradient Boosting: CWB Basics 1



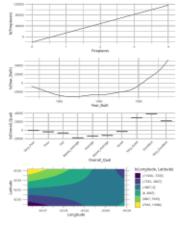
Learning goals

- Concept of CWB
- Which base learners do we use
- Learning goals
- Concept of CWB
 - Which base learners do we use
 - Built-in feature selection



BASE LEARNERS /2

We restrict the base learners to additive model components, i.e.,



linear effect

non-linear (spline) effect

dummy encoded linear model of a cat. feature

tensor product spline for interaction modelling (e.g. spatial effects)

More advanced base learners could also be Markov random fields, random effects, or trees.



BASE LEARNERS /3

Two BLs of the same type can simply be added by adding up their parameter vectors:

$$b_{j}(\mathbf{x}, \boldsymbol{\theta}^{[1]}) + b_{j}(\mathbf{x}, \boldsymbol{\theta}^{[2]}) = b_{j}(\mathbf{x}, \boldsymbol{\theta}^{[1]} + \boldsymbol{\theta}^{[2]}).$$

$$+ \sum_{j=1 \atop j=1 \atop 0.0 \ 2.5 \ 5.0 \ 7.5 \ 10.0}^{\frac{4}{3}} + \sum_{j=1 \atop j=1 \atop 0.0 \ 2.5 \ 5.0 \ 7.5 \ 10.0}^{\frac{4}{3}} = \sum_{j=1 \atop j=1 \atop j=1 \atop 0.0 \ 2.5 \ 5.0 \ 7.5 \ 10.0}^{\frac{4}{3}}$$

Thus, if
$$\{b_j(\mathbf{x}, \boldsymbol{\theta}^{[1]}), b_j(\mathbf{x}, \boldsymbol{\theta}^{[2]})\} \in \mathcal{B}_j$$
, then $b_j(\mathbf{x}, \boldsymbol{\theta}^{[1]} + \boldsymbol{\theta}^{[2]}) \in \mathcal{B}_j$.

