# **Data Mining with Spare Grids**

Seminar: Computational Aspects of Machine Learning

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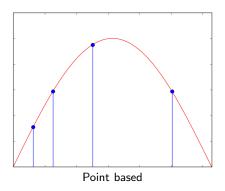
# **Overview**

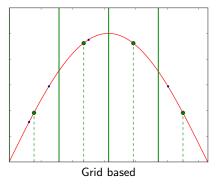
- Motivation for Sparse Grids
- Sparse Grids: Basics
- Sparse Grids: Machine Learning
- Examples with Data Sets
- Parallelization and Implementation

# **Motivation for Sparse Grids**

### Grid based approaches in ML

- Discretizes the space into a grid
- Basis-functions around grid points, not data points





# **Motivation for Sparse Grids**

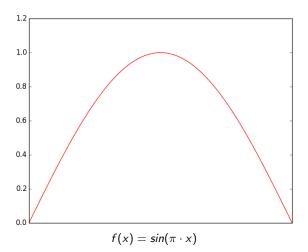
#### Suitable for

- Big datasets
- Easily/automatically classifiable data
- Medical, seismic, commercial data

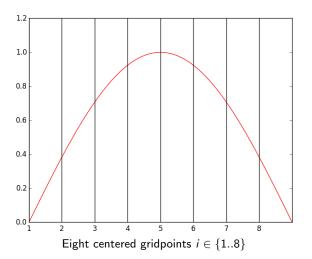
### Curse of dimensionality

- The volume of a space is exponential in it's dimensions
- The amount of training data required becomes unmanageable
  - because of lacking computational/storage capacities
  - because data aquesition is expensive
- Becomes relevant for d > 3
- Applies to full-grid discretization

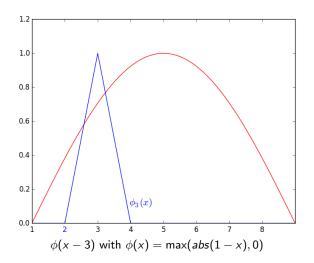
### 1. A function to interpolate



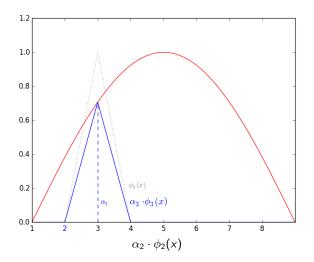
# 2. A (full, regular) grid



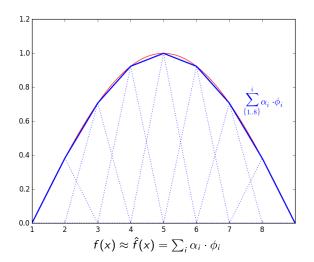
## 3. A basis function (standard hat function)



# 4. A Coefficient (Surplus)



#### Sum over all basis functions



## Full grid interpolation in one dimension

- 1. A function f(x)
- 2. Gridpoints indexed by  $i \in \{1, 2, \dots\}$
- 3. Basis/Ansatz functions; i.e. hat function  $\phi_i(x) = \max(1-|x|,0)$
- 4. Coefficients  $\alpha_i$  (hierarchial surplusses)

$$f(x) \approx \hat{f}(x) = \sum_{i} \alpha_{i} \phi_{i}(x)$$

#### Level of detail

• Level  $l \in \{1, 2, \dots\}$  in addition to index  $i \in 2^l$ 

#### D dimensions

- For each dimension d:  $\hat{f}_d$
- Tensor product over all dimensions  $\hat{f}(x) = \prod_j \hat{f}_j(x)$

# **Sparse Grids** – Basics

#### Hirachial Basis

- Grouping gridpoints into levels  $l_1, l_2, \dots$
- For each level a set of gridpoints  $I_I = \{ i \mid 1 < i < 2^I 1; i \text{ odd } \}$ 
  - $I_1 = \{1\}$
  - $I_2 = \{1, 3\}$
  - $I_3 = \{1, 3, 5, 7\}$
  - ...
- For all dimensions seperately

$$\hat{f}_d(x) = \sum_{l,i} \alpha_{l,i} \cdot \phi_{l,i}(x)$$