# 10.2. Attention Pooling by Similarity

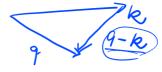
Lecture based on "Dive into Deep Learning" http://D2L.AI (Zhang et al., 2020)

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Similarity Functions

### Attention by similarity



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Nadaraya-Watson estimators rely on a similarity kernel  $\alpha(\mathbf{q},\mathbf{k})$  relating queries  $\mathbf{q}$  to keys  $\mathbf{k}$ . Some common kernels are

$$\alpha(\mathbf{q}, \mathbf{k}) = \exp\left(-\frac{1}{2}\|\mathbf{q} - \mathbf{k}\|^2\right) \qquad \text{Gaussian}$$

$$\alpha(\mathbf{q}, \mathbf{k}) = 1 \text{ if } \|\mathbf{q} - \mathbf{k}\| \le 1 \qquad \text{Boxcar}$$

$$\alpha(\mathbf{q}, \mathbf{k}) = \max\left(0, 1 - \|\mathbf{q} - \mathbf{k}\|\right) \qquad \text{Epanechikov}$$

See for a more extensive review and how the choice of kernels for kernel density estimation.

• All the kernels  $\alpha(\mathbf{k}, \mathbf{q})$  defined here are translation and rotation invariant

i.e., shifting and rotating both k and q in the same way

- Different kernels correspond to different notions of range and smoothness.
  - e.g., the boxcar kernel only attends to observations within a distance of 1.

# **Nadaraya Watson Estimators**

Regression and classification via kernel density estimation

$$\frac{1}{2} = \underline{f(\mathbf{q})} = \sum_{i} \underline{\mathbf{v}}_{i} \underbrace{\frac{\alpha(\mathbf{q}, \mathbf{k}_{i})}{\sum_{j} \alpha(\mathbf{q}, \mathbf{k}_{j})}}.$$

#### For **regression**:

- ullet observations  $(\mathbf{x}_i,y_i)$  (features and labels)
- $\mathbf{v}_i = y_i$  scalars
- $\mathbf{k}_i = \mathbf{x}_i$  are vectors
- query  $\mathbf{q}$  denotes the new location where f should be evaluated.

For (multiclass) **classification**, we use one-hot-encoding of  $\widehat{y_i}$ 

to obtain  $\mathbf{v}_i$ .

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- This estimator requires no training.
- If we narrow the kernel with increasing amounts of data, the approach is consistent, i.e., will converge to some statistically optimal solution.

#### Summary

- Nadaraya-Watson kernel regression is an early precursor of the current attention mechanisms.
- It can be used directly with little to no training or tuning, both for classification and regression.
- The attention weight is assigned according to the similarity (or distance) between query and key.

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