

# Weighted Projection Quantiles Algorithm

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**Algorithm to calculate weighted projection quantile** along the vector  $\mathbf{u} \in \mathbb{R}^p$ , given a set of observations  $\mathbf{X}_1, \mathbf{X}_2, \dots, \mathbf{X}_n$ :

1. **Compute  $Q_{proj}(\mathbf{u})$ , the projection quantile along  $\mathbf{u}$**

- Project each  $\mathbf{X}_i$  along  $\mathbf{u}$  to obtain  $X_{\mathbf{u}i} = \frac{\langle \mathbf{X}_i, \mathbf{u} \rangle}{\|\mathbf{u}\|}$ , for  $i = 1, 2, \dots, n$ .
- Find  $\alpha = \frac{1+\|\mathbf{u}\|}{2}$ -th quantile of  $X_{\mathbf{u}1}, \dots, X_{\mathbf{u}n}$ , say  $q_{\mathbf{u}}$ .
- $Q_{proj}(\mathbf{u}) = q_{\mathbf{u}}\mathbf{e}_{\mathbf{u}}$ ,  $\mathbf{e}_{\mathbf{u}} = \mathbf{u}/\|\mathbf{u}\|$  being the unit vector along  $\mathbf{u}$ .

2. **Compute Weights corresponding to this projection quantile  $Q_{proj}(\mathbf{u})$**

- Compute global weights for the direction vector  $\mathbf{u}$  by  $k$ -mean distance:
  - Compute  $k$ -mean distance corresponding to  $Q_{proj}(\mathbf{u})$  using  $d_k = \frac{1}{n} \sum_{i=1}^n d_i \mathbb{I}_{\{d_i < d_{(k)}\}}$ , where  $d_i$  is the euclidean distance of  $\mathbf{X}_i$  from  $Q_{proj}(\mathbf{u})$  given by  $\|\mathbf{X}_i - Q_{proj}(\mathbf{u})\|$ .  $k$  is a tuning parameter.
  - Compute the weights corresponding to  $\mathbf{u}$ :

$$w_{\mathbf{u}} = \exp(-a.d_k)$$

where  $a$  is a tuning parameter.

- Compute weights for each sample point  $\mathbf{X}_i; i = 1, 2, \dots, n$ :
  - Compute the orthogonal Norms by  $\|\mathbf{X}_{\mathbf{u}\perp i}\| = \|\mathbf{X}_i - X_{\mathbf{u}i}\mathbf{e}_{\mathbf{u}}\|$ .
  - Compute weight of  $i^{th}$  sample:

$$w_{2i} = \begin{cases} \exp \left[ -b \frac{\|\mathbf{X}_{\mathbf{u}\perp i}\|}{\|\mathbf{X}_i\|} \right] & \text{if } \|\mathbf{X}_i\| \leq \epsilon \\ 0 & \text{otherwise} \end{cases}$$

$b, \epsilon$  being tuning parameters.

3. **Compute the weighted projection quantile**

- Define  $\tilde{X}_{\mathbf{u}i} = w_{\mathbf{u}}w_{2i}X_{\mathbf{u}i}$  for  $i = 1, \dots, n$ .
- Find  $\alpha = \frac{1+\|\mathbf{u}\|}{2}$  th quantile of  $\tilde{X}_{\mathbf{u}1}, \dots, \tilde{X}_{\mathbf{u}n}$ . Let it be  $\tilde{q}_{\mathbf{u}}$ .
- Find the weighted projection quantile as  $\tilde{Q}_{proj}(\mathbf{u}) = \tilde{q}_{\mathbf{u}}\mathbf{e}_{\mathbf{u}}$ .

***Algorithm for weighted projection quantile based data depth given data  $\mathbf{X}_1, \dots, \mathbf{X}_n$  and a new point  $\mathbf{p}$***

1. Compute Projection Quantiles for all  $n$  sample points along the direction of  $\mathbf{p}$ .
2. Find two closest points along the grid and their corresponding quantile values, say  $\alpha_1$  and  $\alpha_2$  corresponding to  $u_{p1}$  and  $u_{p2}$ .
3. Approximate the quantile value for  $p$  by interpolation  $\alpha_p = \frac{(\alpha_1 - \alpha_2)(\|p\| - \|u_1\|)}{\|u_1\| - \|u_2\|} + \alpha_1$ .
4. Define **projection quantile depth** as  $d_p = \exp(-\alpha_p)$ .

***Algorithm for weighted projection depth based classification scheme***

1. Given data with  $k$  classes and a new observation  $p$  find the data depth for  $p$  for each class  $k$ .
2. Classify  $p$  in class  $j$  if  $d_{pj} = \min\{d_{pi}; i \in \{1, \dots, k\}\}$