

# Kernel Probability Density Function

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Kernel smoothers: create a continuous function to represent the similarities between a neighborhood of points.  
Bandwith: a scaling factor used to define the width of probability mass around a point.

## 1 Kernels and Kernel Distributions

### 1.1 Kernel Smoothing Function

Smoothing functions: represent points via a set of adjusting weights and base points:

$$\hat{f}(x) = \sum_{i=1}^n W_i(x) y_i$$

Kernel smoothing function does this in terms of a scaling factor (bandwidth), and makes the weights in terms of a continuous function with total density = 1.

$$W_{hi}(x) = \frac{K\left(\frac{|x_0 - x_i|}{\lambda}\right)}{\sum_{i=1}^n K\left(\frac{|x_0 - x_i|}{\lambda}\right)}$$

Kernel smoother thus is:

$$\hat{f}(x) = \sum_{i=1}^n W_{hi}(x) y_i$$

### 1.2 Kernel Probability Distribution Function

The kernel distribution function is as follows, with  $N(x_0)$  being the neighborhood of total points in the prob distribution,  $\lambda$  being the bandwidth of the function, indicating the scale factor of the distribution.

$$\hat{f}(x_0) = \frac{1}{N\lambda} \sum_{i=1}^N K_{\lambda}(x_0, x_i)$$

The Kernel function can be one of many different distributions:

$$\textit{Epanechnikov} : D(t) = \frac{3}{4}(1 - t^2) \text{ such that } |t| < 1$$

$$\textit{Gaussian} : D(t) = \frac{1}{\sqrt{2\pi}}(e^{-\frac{1}{2}t^2})$$

$$\textit{Tri - Cube} : D(t) = (1 - t^3)^3 \text{ such that } |t| < 1$$

These are only a couple potential equations of base kernels. Other examples include triangle and box. These distributions are implemented such that:

$$K_\lambda = D\left(\frac{x_0 - x_i}{\lambda}\right)$$

### 1.3 Bandwidth

The bandwidth is a scaling factor used to define the width of probability mass around a point. Increasing bandwidth increases bias and decreases variance.

## 2 Pseudocode of Kernel Probabilities

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**Algorithm 1** kernelprob: Generate a probability estimate based on a distribution with a kernel base function.

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**Input:**  $x_0, X = x_0, x_1, \dots, x_n$ , bandwidth  $\lambda$ , kernel function  $K_\lambda$

**Output:** kernel based probability of  $x_0$  in  $X$

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1: procedure KERNELPROB( $\mathcal{X}, \lambda, K_\lambda$ )
2:    $kernsum = 0$ ;
3:   for  $i$  in  $range(0, n)$  do                                 $\triangleright$  run kernel func on each point
4:      $kernsum += K\left(\frac{|x_i - x_0|}{\lambda}\right)$                  $\triangleright$  get kernel dist between 2 points
5:   return  $\frac{1}{n\lambda} \times kernsum$ 

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## 3 Pseudocode of Different Kernel Base Distributions

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**Algorithm 2** epanechnikov: Epanechnikov distribution used to calculate distance between 2 points.

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**Input:** Value to verify  $t$

**Output:** Distance  $d$

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1: procedure EPANECHKOV( $t$ )
2:   if  $|t| < 1$  then
3:     return  $\frac{3}{4} \times (1 - t^2)$ 
4:   else
5:     return 0

```

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**Algorithm 3** gaussian: Gaussian distribution used to calculate distance between 2 points.

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**Input:** Value to verify  $t$

**Output:** Distance  $d$

```

1: procedure GAUSSIAN( $t$ )
2:   return  $\frac{1}{\sqrt{2\pi}}(e^{-\frac{1}{2}t^2})$ 

```

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**Algorithm 4** tricube: Tri-cube distribution used to calculate distance between 2 points.

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**Input:** Value to verify  $t$

**Output:** Distance  $d$

```
1: procedure TRICUBE( $t$ )  
2:   if  $|t| < 1$  then  
3:     return  $(1 - t^3)^3$   
4:   else  
5:     return 0
```

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