

# Master of Technology

## U2/6: Computational Intelligence I

### To clarify slide 25

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### Slide 25 in lecture note

Input space  $\vec{x}_i = (x_{i1}, x_{i2})$  Feature space  $\vec{X}_i = (x_{i1}^2, x_{i2}^2, \sqrt{2}x_{i1}x_{i2})$

Inner product  $\vec{X}_i \cdot \vec{X}_j = X_{i1}X_{j1} + X_{i2}X_{j2} + X_{i3}X_{j3}$

To get the new first dimension: 1 multiplication

Second dimension: 1 multiplication

Third dimension: 2 multiplications

In all,  $1+1+2 = 4$  multiplications.

Multiplications: 8 (for the projections) + 3 (in the dot product) = 11 multiplications

Additions: 2 (in the dot product)

**Total:  $11 + 2 = 13$  operations.**

Without kernel trick

With kernel trick

$$\begin{aligned} K(\vec{x}_i, \vec{x}_j) &= (\vec{x}_i \cdot \vec{x}_j)^2 \\ &= (x_{i1}x_{j1} + x_{i2}x_{j2})^2 \\ &= x_{i1}^2x_{j1}^2 + x_{i2}^2x_{j2}^2 + 2x_{i1}x_{i2}x_{j1}x_{j2} \\ &= (x_{i1}^2, x_{i2}^2, \sqrt{2}x_{i1}x_{i2}) \cdot (x_{j1}^2, x_{j2}^2, \sqrt{2}x_{j1}x_{j2}) \end{aligned}$$

Multiplications: 2 (for the dot product in the original space) + 1 (for squaring the result) = 3 multiplications

Additions: 1 (for the dot product in the original space)

**Total:  $3 + 1 = 4$  operations.**

Source: <https://blog.statsbot.co/support-vector-machines-tutorial-c1618e635e93>

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## Recall: trained SVM model (Slide 22)

Decision plane  $d(\mathbf{x}) = \mathbf{w}^T \phi(\mathbf{x}) + b$

Provided in  
sample data

$$d(\mathbf{x}) = \sum_{i=1}^l y_i \alpha_i \langle \phi(\mathbf{x}), \phi(\mathbf{x}_i) \rangle + b$$

Parameters associated  
with input vectors in  
sample data,  $\alpha_i \neq 0$  for  
support vectors

All the information the learning algorithm needs is the **inner products** between data points in the feature space, where  $\mathbf{x}, \mathbf{x}_i$  ( $i = 1, \dots, l$ )  $\in \mathbf{X}$ , the input space

## Dot product in (old) 2-D input space

Input space  $\vec{x}_i = (x_{i1}, x_{i2})$

$$\begin{aligned}
 K(\vec{x}_i, \vec{x}_j) &= (\vec{x}_i \cdot \vec{x}_j) \\
 &= (x_{i1}x_{j1} + x_{i2}x_{j2})
 \end{aligned}$$

## Dot product in (new) 3-D feature space

Input space  $\vec{x}_i = (x_{i1}, x_{i2})$     Feature space  $\vec{X}_i = (x_{i1}^2, x_{i2}^2, \sqrt{2}x_{i1}x_{i2})$

$$\begin{aligned} K(\vec{x}_i, \vec{x}_j) &= (\vec{x}_i \cdot \vec{x}_j)^2 \\ &= (x_{i1}^2, x_{i2}^2, \sqrt{2}x_{i1}x_{i2}) \cdot (x_{j1}^2, x_{j2}^2, \sqrt{2}x_{j1}x_{j2}) \\ &= x_{i1}^2x_{j1}^2 + x_{i2}^2x_{j2}^2 + 2x_{i1}x_{i2}x_{j1}x_{j2} \end{aligned}$$

## Solution 1: Transform data, calculate dot product

Input space  $\vec{x}_i = (x_{i1}, x_{i2})$     Feature space  $\vec{X}_i = (x_{i1}^2, x_{i2}^2, \sqrt{2}x_{i1}x_{i2})$

Inner product  $\vec{X}_i \cdot \vec{X}_j = X_{i1}X_{j1} + X_{i2}X_{j2} + X_{i3}X_{j3}$

To **map EACH data** point from old space to new space

To get the new first dimension: 1 multiplication

Second dimension: 1 multiplication

Third dimension: 2 multiplications

In all,  $1+1+2 = 4$  multiplications.

To **calculate the dot product in (new) feature space**

Multiplications: 8 (for the projections) + 3 (in the dot product) = 11 multiplications

Additions: 2 (in the dot product)

**Total:  $11 + 2 = 13$  operations.**

## Solution 2: Trick

### Dot product in new space

$$\begin{aligned}
 & (x_{i1}^2, x_{i2}^2, \sqrt{2}x_{i1}x_{i2}) \cdot (x_{j1}^2, x_{j2}^2, \sqrt{2}x_{j1}x_{j2}) \\
 &= x_{i1}^2 x_{j1}^2 + x_{i2}^2 x_{j2}^2 + 2x_{i1}x_{i2}x_{j1}x_{j2} \qquad \qquad \qquad = (x_{i1}x_{j1} + x_{i2}x_{j2})^2
 \end{aligned}$$

### Dot product in old space

## Solution 2: Transform dot product

Input space  $\vec{x}_i = (x_{i1}, x_{i2})$     Feature space  $\vec{X}_i = (x_{i1}^2, x_{i2}^2, \sqrt{2}x_{i1}x_{i2})$

Inner product  $\vec{X}_i \cdot \vec{X}_j = X_{i1}X_{j1} + X_{i2}X_{j2} + X_{i3}X_{j3}$

$$\begin{aligned}
 & (x_{i1}^2, x_{i2}^2, \sqrt{2}x_{i1}x_{i2}) \cdot (x_{j1}^2, x_{j2}^2, \sqrt{2}x_{j1}x_{j2}) \\
 &= x_{i1}^2 x_{j1}^2 + x_{i2}^2 x_{j2}^2 + 2x_{i1}x_{i2}x_{j1}x_{j2} \qquad \qquad \qquad = (x_{i1}x_{j1} + x_{i2}x_{j2})^2
 \end{aligned}$$

- 2 multiplications (for the dot product in the original space)
- 1 addition (for the dot product in the original space)
- 1 (for squaring the result)

Total: 3 + 1 = 4 operations.

## Comparison between two solutions

Input space  $\vec{x}_i = (x_{i1}, x_{i2})$  Feature space  $\vec{X}_i = (x_{i1}^2, x_{i2}^2, \sqrt{2}x_{i1}x_{i2})$

Inner product  $\vec{X}_i \cdot \vec{X}_j = X_{i1}X_{j1} + X_{i2}X_{j2} + X_{i3}X_{j3}$

To get the new first dimension: 1 multiplication

Second dimension: 1 multiplication

Third dimension: 2 multiplications

In all,  $1+1+2 = 4$  multiplications.

Multiplications: 8 (for the projections) + 3 (in the dot product) = 11 multiplications

Additions: 2 (in the dot product)

**Total:  $11 + 2 = 13$  operations.**

**Solution 1**

**Solution 2**

$$\begin{aligned} K(\vec{x}_i, \vec{x}_j) &= (\vec{x}_i \cdot \vec{x}_j)^2 \\ &= (x_{i1}x_{j1} + x_{i2}x_{j2})^2 \\ &= x_{i1}^2x_{j1}^2 + x_{i2}^2x_{j2}^2 + 2x_{i1}x_{i2}x_{j1}x_{j2} \\ &= (x_{i1}^2, x_{i2}^2, \sqrt{2}x_{i1}x_{i2}) \cdot (x_{j1}^2, x_{j2}^2, \sqrt{2}x_{j1}x_{j2}) \end{aligned}$$

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Multiplications: 2 (for the dot product in the original space) + 1 (for squaring the result) = 3 multiplications

Additions: 1 (for the dot product in the original space)

**Total:  $3 + 1 = 4$  operations.**

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## Examples of Kernel functions

- **Polynomial** kernel with degree  $d$

$$K(\mathbf{x}, \mathbf{y}) = (\mathbf{x}^T \mathbf{y} + 1)^d$$

- **Radial basis function** kernel with width  $\sigma$

$$K(\mathbf{x}, \mathbf{y}) = \exp(-\|\mathbf{x} - \mathbf{y}\|^2 / (2\sigma^2))$$

