NMK40403 ARTIFICIAL INTELLIGENCE

SUPPORT VECTOR MACHINES (B)

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- Given two vectors $u(u_1, u_2)$ and $v(v_1, v_2)$ then : $u+v=(u_1+v_1u_2+v_2)$
- *Which means that adding two vectors gives us a third vector whose coordinate are the sum of the coordinates of the original vectors.*

• Example:

The difference between two vectors

The difference works the same way:

u-v=(u₁-v₁,u₂-v₂)

Since the subtraction is not commutative, we can also consider the other case:

$$
v-u=(v_1-u_1,v_2-u_2)
$$

- The last two pictures describe the "true" vectors generated by the difference of u and v.
- However, since a vector has a magnitude and a direction, we often consider that parallel translate of a given vector (vectors with the same magnitude and direction but with a different origin) are the same vector, just drawn in a different place in space.
- So don't be surprised if you meet the following (Figure 17 (a) & (b) :

Figure 17 (a) & (b) : another way to view the difference v-u

• If you do the math, it looks wrong, because the end of the vector u−v is not in the right point, but it is a convenient way of thinking about vectors which you'll encounter often.

The dot product

- One very important notion to understand SVM is the dot product.
- Definition:

Geometrically, it is the product of the Euclidian magnitudes of the two vectors and the cosine of the angle between them

• Which means if we have two vectors \overline{x} and \overline{y} and there is an angle θ (theta) between them, their dot product is:

 $x \cdot y = ||x|| ||y|| \cos(\theta)$

Why ?

• To understand let's look at the problem geometrically.

- In the definition, they talk about $cos(\theta)$, let's see what it is.
- By definition we know that in a right-angled triangle: $cos(\theta) =$ adjacent hypotenuse
- In our example, we don't have a right-angled triangle.
- However if we take a different look Figure 18 we can find two right-angled triangles formed by each vector with the horizontal axis.

Figure 19 (a) & (b)

• So now we can view our original schema like this:

• We can see that

 $\theta = \beta - \alpha$

- So computing $cos(\theta)$ is like computing $cos(\beta-\alpha)$
- There is a special formula called the difference identity for cosine which says that:

 $cos(\beta-\alpha)=cos(\beta)cos(\alpha)+sin(\beta)sin(\alpha)$

• So if we replace each term

 $cos(\theta)=cos(\beta-\alpha)=cos(\beta)cos(\alpha)+sin(\beta)sin(\alpha)$ $cos(\theta) = \frac{x_1}{\ln x}$ \mathcal{X} y_1 $\frac{y_1}{y} + \frac{x_2}{\|x\|}$ \mathcal{X} y_2 \mathcal{Y} $\cos(\theta) = \frac{x_1 y_1 + x_2 y_2}{\|x\| \|\|x\|}$ x || || y

- If we multiply both sides by ||x||||y|| we get: $\|x\| \|\nu\| \cos(\theta) = x_1 y_1 + x_2 y_2$
- Which is the same as: $\|x\|$ |v $\|cos(\theta)=x\cdot v$
- This is the geometric definition of the dot product !

- Eventually from the two last equations we can see that : $x \cdot y = x_1 y_1 + x_2 y_2 = \sum_{i=1}^{2} (x i y i)$
- This is the algebraic definition of the dot product !

A few words on notation

- The dot product is called like that because we write a dot between the two vectors.
- Talking about the dot product x⋅y is the same as talking about
	- the *inner product* $\langle x,y \rangle$ (in linear algebra)
	- *scalar product* because we take the product of two vectors and it returns a scalar (a real number)

SVM - Understanding the math - Part 1 - The margin The orthogonal projection of a vector

• Given two vectors x and y, we would like to find the orthogonal projection of x onto y.

SVM - Understanding the math - Part 1 - The margin • To do this we project the vector x onto y

Figure 22

SVM - Understanding the math - Part 1 - The margin • This give us the vector z

Figure 23 : z is the projection of x onto y

• By definition :

 $cos(\theta) = \frac{||z||}{||w||}$ \mathcal{X} $\|z\|=\|x\|\cos(\theta)$

- We saw in the section about the dot product that $cos(\theta) = \frac{x.y}{\ln x \ln x}$ x || || y
- So we replace $cos(\theta)$ in our equation: $||z|| = ||x|| \frac{x.y}{||x|| ||u||}$ $x||$ ||y $\|z\| = \frac{x \cdot y}{\|x\|}$ \mathcal{Y}

- If we define the vector u as the direction of y then $u=\frac{y}{\ln x}$ $\frac{y}{y\parallel}$ and ∥z∥=u.x
- We now have a simple way to compute the norm of the vector z.
- Since this vector is in the same direction as y it has the direction u

 $u=\frac{z}{\ln z}$ Z $z=$ $||z||$ u

• And we can say :

The vector z=(u⋅*x)u is the orthogonal projection of x onto y.*

SVM - Understanding the math - Part 1 - The margin Why are we interested by the orthogonal projection ? Well in our example, it allows us to compute the distance between x and the line which goes through y.

The SVM hyperplane

Understanding the equation ofthe hyperplane

- You probably learnt that an equation of a line is : $v=ax+b$. However when reading about hyperplane, you will often find that the equation of an hyperplane is defined by : *w^Tx=0*
- How does these two forms relate?
- In the hyperplane equation you can see that the name of the variables are in bold. Which means that they are vectors ! Moreover, *w^Tx* is how we compute the inner product of two vectors, and if you recall, the inner product is just another name for the dot product !

- Note that $y = ax + b$
- is the same thing as y−ax−b=0
- Given two vectors w $-b$ $-a$ 1 and x 1 χ \mathcal{Y}

$$
w^{T}x = -b \times (1) + (-a) \times x + 1 \times y
$$

$$
w^{T}x = y - ax - b
$$

- The two equations are just different ways of expressing the same thing.
- It is interesting to note that w_0 is -b, which means that this value determines the intersection of the line with the vertical axis.

• Why do we use the hyperplane equation $w^T x$ instead of $v = ax + b$?

For two reasons:

- it is easier to work in more than two dimensions with this notation,
- the vector w will always be normal to the hyperplane (Note: w will always be normal because we use this vector to define the hyperplane, so by definition it will be normal. As you can see, when we define a hyperplane, we suppose that we have a vector that is orthogonal to the hyperplane)
- And this last property will come in handy to compute the distance from a point to the hyperplane.

SVM - Understanding the math - Part 1 - The margin Compute the distance from a point to the hyperplane In Figure 25 we have an hyperplane, which separates two group of data. \overline{D}

- To simplify this example, we have set $w_0=0$.
- As you can see on the Figure 25, the equation of the hyperplane is:

$$
x_2 = -2x_1
$$

- which is equivalent to W^T γ
- with $w\binom{2}{1}$ 1 and x x_1 x_2
- Note that the vector w is shown on the Figure 25. (w is not a data point)

- We would like to compute the distance between the point A(3,4) and the hyperplane.
- This is the distance between A and its projection onto the hyperplane

- We can view the point A as a vector from the origin to A.
- If we project it onto the normal vector w

• We get the vector p

- Our goal is to find the distance between the point A(3,4) and the hyperplane. We can see in Figure 28 that this distance is the same thing as ∥p∥.
- Let's compute this value.
- We start with two vectors, $w=(2,1)$ which is normal to the hyperplane, and $a=(3,4)$ which is the vector between the origin and A.

$$
||w|| = \sqrt{2^2 + 1^2} = \sqrt{5}
$$

Let the vector u be the direction of w

$$
u = \left(\frac{2}{\sqrt{5}}, \frac{1}{\sqrt{5}}\right)
$$

• p isthe orthogonal projection of a onto w so : $n = (n \ a)$

$$
p = \left(3 \times \frac{2}{\sqrt{5}} + 4 \times \frac{1}{\sqrt{5}}\right)u
$$

$$
p = \left(\frac{6}{\sqrt{5}} + \frac{4}{\sqrt{5}}\right)u
$$

$$
p = \frac{10}{\sqrt{5}}u
$$

$$
p = \left(\frac{10}{\sqrt{5}} \times \frac{2}{\sqrt{5}} \times \frac{10}{\sqrt{5}}\right)
$$

$$
p = \left(\frac{20}{\sqrt{5}} \times \frac{10}{\sqrt{5}}\right)
$$

$$
p = \left(\frac{20}{\sqrt{5}} \times \frac{10}{\sqrt{5}}\right)
$$

$$
p = (4,2)
$$

$$
||p|| = \sqrt{4^2 + 2^2} = 2\sqrt{5}
$$

Compute the margin ofthe hyperplane

• Now that we have the distance ∥p∥ between A and the hyperplane, the margin is defined by :

$$
margin = 2||p|| = 4\sqrt{5}
$$

• We did it ! We computed the margin of the hyperplane !

