

Module 2

Inner products of functions and random (variables) variable

In previous session we looked at properties of inner products, to compute length, angle and distance

We focused on inner products of finite dimensional vector space.

In this session we will look at 2 examples of inner product of other types of vectors
i.e inner product of functions and inner product of random variables

The inner products we discussed so far we defined for vectors with finite number of entries, and we can think of these vectors as discrete functions with finite number of function values.

The Concept of inner product can be generalized to Continuous valued functions as well.

And then the sum of individual component of vectors turns into an integral.

The inner product between two functions is defined as follows:

"The inner product between two functions, u and v , is the integral of the interval from a to b , of u of x , times v of x , dx "

$$\langle u, v \rangle = \int_a^b u(x)v(x) dx.$$

And as with our normal inner product, we can define norms and orthogonality by looking at this inner product.

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if that integral evaluates to 0, the functions
 u and v are orthogonal

Let's have look at an example,

"If we choose, u of x equals $\sin x$, and v of x is
 \cos of x , and we define f of x to be u of x ,
times v of x

$$\therefore u(x) = \sin(x)$$

$$v(x) = \cos(x)$$

$$f(x) = u(x)v(x)$$

then we going to end up with this function

this function is $\sin(x)$ times $\cos(x)$ (see pc)

we see that this function is odd, which means that $f(-x) = -f(x)$.

If we choose the integral limit, to be minus π and plus π , then the integral of this product $\sin(x) \times \cos(x)$, evaluates to 0.

That means that \sin and \cos are orthogonal.

And actually holds, that if you "look at set of functions say $1, \cos x, \cos 2x, \cos 3x$ and so on, that all of these functions are orthogonal to each other if we integrate from $-\pi$ to $+\pi$ "

$$\therefore \{1, \cos x, \cos 2x, \cos 3x, \dots\}$$

Another example for defining an inner product, between unusual types are random variables or random vectors.

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If we have 2 random variables which are uncorrelated, then we know that the following relationship.

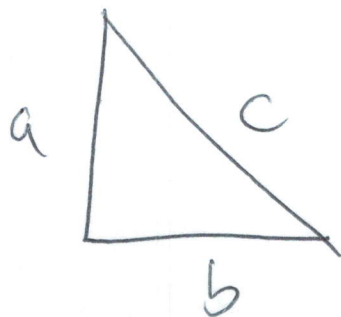
"We know that the variance of $x+y$ is variance of x , plus variance of y , where x and y are random variables"

$$\therefore \text{Var}[x+y] = \text{var}[x] + \text{var}[y]$$

If we remember that variance are measured in squared units, this looks very much like the Pythagorean theorem for right triangle

Let me state that " c squared equals a squared plus b squared", if we look at triangle of this form

$$c^2 = a^2 + b^2$$



Let's see if we can find a geometric interpretation of
variance relation of uncorrelated variables.

Random variables can be considered
vectors in vector space if we can define
inner products to obtain geometric
properties of the these random variables

"We define the inner product between 2 random
variables, between x and y , to be the covariance
between x, y

$$\langle x, y \rangle = \text{Cov}[x, y]$$

We see that the covariance is

- symmetric
- positive definite, and
- linear

So (linear) linearity would mean:

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hat "Covariance of 1 times x plus y and z ,
where x, y , and z are random variable, and
 1 is real number, is 1 times Covariance between
 x and z plus Covariance between y and z ."

$$\text{Cov}(1x+y, z) = 1\text{Cov}[x, z] + \text{Cov}[y, z]$$

And if "length of random variable is square root of
Covariance between x with itself, which is square root
of variance of x , this is standard deviation of
random variable x ."

$$\therefore \sqrt{\text{Cov}[x, x]} = \sqrt{\text{Var}[x]} = \sigma(x)$$

"This is the length of random variable"

$$\|x\| = \sqrt{\text{Cov}[x, x]} = \sqrt{\text{Var}[x]} = \sigma(x)$$

Therefore the zero(0) vector is a vector that has no uncertainty, that means standard deviation is 0.

If we now look at the angle, between random variables, we get the following relationship.

We get "the Cos of θ , which is the angle between two random variables, is by definition the inner product between the two random variables divided by the length of first random variable times length of second random variable

$$\cos \theta = \frac{\langle x, y \rangle}{\|x\| \|y\|}$$

We can now write this out using the definition of our inner product,

i.e $\langle x, y \rangle = \text{Cov}[x, y]$

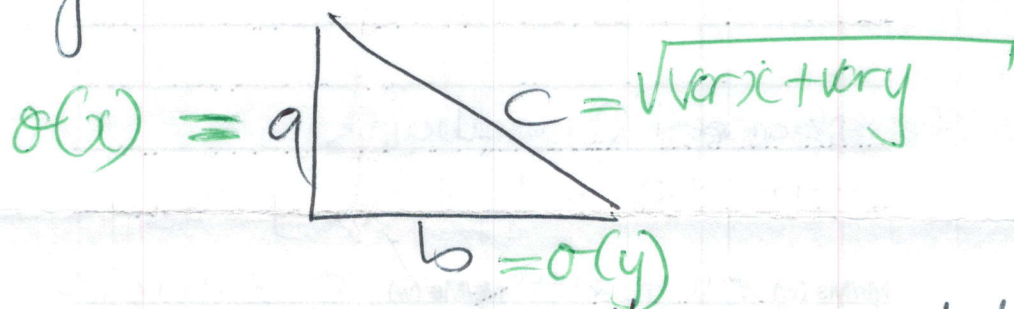
We get "Covariance between x and y divided by square root of variance of x , times variance of y .

$$= \frac{\text{cov}[x, y]}{\sqrt{\text{var}[x] \text{var}[y]}}$$

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and this evaluates to 0, if and only if, the Covariance between x and y is 0, and that's the case when x and y are uncorrelated.

Coming back now to our geometric interpretation.



$\sigma(x) = a$

$c = \sqrt{\text{var}[x] + \text{var}[y]}$

$b = \sigma(y)$

"we will now replace a with standard deviation (σ) of x , b is std dev of y , and c is square root of variance of x + variance of y "

and this is how we get our geometric interpretation of random variables

In this session, we looked at inner products of rather unusual objects, functions and random variables.

However, even with functions and random variables, the inner product allows us to think about length and angles between these objects.

In case of random variables we saw that, the variance of sum of two uncorrelated random variables, can be geometrically interpreted using the Pythagorean theorem.