## Normal distributions on vector spaces

#### Kristian Wichmann

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### 1 Affine transformations of euclidean spaces

Let  $s: \mathbb{R}^n \to \mathbb{R}^m$  be a linear transformation. This means that the is an  $m \times n$  matrix A so s(x) = Ax.

An *affine* transformation t is formed by following this linear map by a translation:

$$t: \mathbb{R}^n \to \mathbb{R}^m, t(x) = Ax + v \tag{1.1}$$

Here,  $v \in \mathbb{R}^m$ . Since translations are always bijective, we note that t is bijective iff A is invertible.

Each component of an affine transformation is composed from measurable function - is is understood that we mean with respect to the Borel algebras of each space) - so the affine transformation itself is measurable as well.

# 1.1 Transformation properties of the Lebesgue measure

Recall that the Lebesgue measure in n dimensions  $m_n$  is invariant under translation: If t is a translation  $t: \mathbb{R}^n \to \mathbb{R}^n, t(x) = x + x_0$ , where  $x_0 \in \mathbb{R}^n$  then:

$$t(m_n) = m_n \tag{1.2}$$

Also, if  $s: \mathbb{R}^n \to \mathbb{R}^n, x \mapsto Ax$  is an isomorphism, then:

$$s(m_n) = m_n |\det A^{-1}|$$
 (1.3)

Combining the two, the formula for affine transformation is the same as for linear ones.

#### 2 Orthogonal complement

Let V be a finite-dimensional vector space with an inner product  $\langle \cdot, \cdot \rangle$ . Let U be a subspace of V. Then we define the *orthogonal complement* of U as:

$$U^{\perp} = \{ v \in V | \forall u \in U : \langle u, v \rangle = 0 \}$$
 (2.1)

**Theorem 2.1.**  $U^{\perp}$  is a subspace of V.

*Proof.* According to the subspace theorem, we need to show three things:

- $U^{\perp}$  is not empty: Clearly  $0 \in U^{\perp}$ .
- Closed under addition: If  $v_1, v_2 \in U^{\perp}$ , then for all  $u \in U^{\perp}$ :

$$\langle v_1 + v_2, u \rangle = \langle v_1, u \rangle + \langle v_2, u \rangle = 0 \tag{2.2}$$

• Closed under scalar multiplication: If  $v \in U^{\perp}$  and  $c \in \mathbb{R}$  then for all  $u \in U^{\perp}$ :

$$\langle cv, u \rangle = c \langle v, u \rangle = 0$$
 (2.3)

Since the only vector perpendicular to itself is 0, we further conclude that  $U \cap U^{\perp} = \{0\}.$ 

**Theorem 2.2.** If  $e_1, e_2, \ldots, e_m$  is an orthonormal basis for U, then for any  $v \in V$ :

$$v - \sum_{i=1}^{m} \langle v, e_i \rangle e_i \in U^{\perp} \tag{2.4}$$

*Proof.* Let  $u \in U$ . Then we can write  $u = \sum_{j=1}^{m} \lambda_j e_j$  for some coefficients  $\lambda_j$ . Now calculate the inner product with the vector above:

$$\langle v - \sum_{i=1}^{m} \langle v, e_i \rangle e_i , \sum_{j=1}^{m} \lambda_j e_j \rangle = \sum_{i=j}^{m} \lambda_j \langle v, e_j \rangle - \sum_{i=1}^{m} \sum_{j=1}^{m} \lambda_j \langle v, e_i \rangle \langle e_i, e_j \rangle$$
 (2.5)

Since  $\langle e_i, e_j \rangle = \delta_{ij}$  this vanishes.

This means that we may write any  $v \in V$  as a sum of vectors from U and  $U^{\perp}$  respectively:

$$v = \underbrace{\sum_{i=1}^{m} \langle v, e_i \rangle e_i}_{\in U} + \underbrace{v - \sum_{i=1}^{m} \langle v, e_i \rangle e_i}_{\in U^{\perp}}$$
(2.6)

**Theorem 2.3.** The decomposition into elements from U and  $U^{\perp}$  from equation 2.6 is unique.

*Proof.* Let  $v = u_1 + u_1^{\perp}$  and  $v = u_2 + u_2^{\perp}$  be two such decompositions. Then  $u_1 + u_1^{\perp} = u_2 + u_2^{\perp}$  and hence  $u_1 - u_2 = u_2^{\perp} - u_1^{\perp}$ . But this means that this vector is a member of both U and  $U^{\perp}$ , and hence it must be 0. This means  $u_1 = u_2$  and  $u_1^{\perp} = u_2^{\perp}$ .

#### 2.1 The orthogonal projection

The previous section motivates the following:

**Definition 2.1.** Let V be a finite-dimensional inner product vector space and U a subspace of V. The orthogonal projection from V onto U is the map  $p:V \to V$  which satisfies:

$$\forall v \in V: \quad p(v) \in U, \quad v - p(v) \in U^{\perp} \tag{2.7}$$

As we see, one could also define the co-domain of p to be U. Usually, the distinction will not matter much.

**Theorem 2.4.** The orthogonal projection operator is linear.

*Proof.* We need to show additivity and homogeneity:

• Additivity: Let  $v_1, v_2 \in V$ . Then  $p(v_1) + p(v_2) \in U$  and:

$$v_1 - p(v_1) + v_2 - p(v_2) = v_1 + v_2 - (p(v_1) + p(v_2)) \in U^{\perp}$$
 (2.8)

Adding the two we get  $v_1 + v_2$ . So  $p(v_1 + v_2) = p(v_1) + p(v_2)$ .

• Homogeneity. Let  $v \in V$  and  $c \in \mathbb{R}$ . Then  $cp(v) \in U$  and  $c(v - p(v)) = cv - cp(v) \in U^{\perp}$ . Adding the two we get cv, so p(cv) = cp(v).

**Theorem 2.5.** The orthogonal projection operator  $p: V \to V$  is idempotent. *I.e.*  $p \circ p = p$ .

*Proof.* Let  $v \in V$ . Then  $p(v) \in U$ . But this means that the decomposition of p(v) is p(v) + 0. So  $p \circ p(v) = p(v)$ .

#### 3 Lebesgue measures on vector spaces

#### 3.1 Coordinate maps

Let V be a finite-dimensional vector space of dimension n. Our question is, if we can turn V into a measure space in a natural way. Since we know that V is isomorphic to  $\mathbb{R}^n$ , it makes sense to tweak the usual Lebesgue measure in N dimensions:

Let  $e_1, e_2, \ldots, e_n$  be a basis for V. Then we can define the *coordinate map* as follows:

$$\phi: \mathbb{R}^n \to V, \quad \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} \mapsto \sum_{i=1}^n x_i e_i$$
 (3.1)

This is obviously an isomorphism. Specifically, it is invertible with inverse  $\phi^{-1}: V \to \mathbb{R}^n$ .

The coordinate map depends on the chosen basis. If we had chosen another basis  $e_1^*, e_2^*, \ldots, e_n^*$  we would get another isomorphism  $\phi^*$ .

#### 3.2 Borel algebra on V

We can now use  $\phi^{-1}$  to induce a  $\sigma$ -algebra on V. Set  $\mathbb{B}_V$  to the smallest  $\sigma$ -algebra that makes  $\phi^{-1}$  measurable when  $\mathbb{R}^n$  is equipped with the Borel algebra  $\mathbb{B}_n$ . We call  $\mathbb{B}_V$  the Borel algebra on V.

At first this object seems to depend of the choice of basis for V. But it turns out that the use of definite article in the definition is justified:

**Theorem 3.1.** If  $e_1, e_2, \ldots, e_n$  and  $e_1^*, e_2^*, \ldots, e_n^*$  are bases for V, then the induced  $\sigma$ -algebra  $\mathbb{B}_V$  and  $\mathbb{B}_V^*$  is the same thing.

*Proof.* We know that  $\phi^{-1}$  is  $\mathbb{B}_V - \mathbb{B}_n$  measurable by definition. We have:

$$(\phi^*)^{-1} = (\phi^*)^{-1} \circ \mathrm{id}_V = (\phi^*)^{-1} \circ (\phi \circ \phi^{-1}) = ((\phi^*)^{-1} \circ \phi) \circ \phi^{-1}$$
 (3.2)

 $((\phi^*)^{-1} \circ \phi)$  is a linear operator on  $\mathbb{R}^n$  and so according to section 1.1 is measurable. So  $(\phi^*)^{-1}$  must be  $\mathbb{B}_V - \mathbb{B}_n$ -measurable. Since  $\mathbb{B}_V^*$  is the smallest  $\sigma$ -algebra to make  $(\phi^*)^{-1} \mathbb{B}_V - \mathbb{B}_n$ -measurable, we must have  $\mathbb{B}_V^* \subseteq \mathbb{B}_V$ .

But by a totally symmetric argument, we must also have  $\mathbb{B}_V \subseteq \mathbb{B}_V^*$ . Hence  $\mathbb{B}_V = \mathbb{B}_V^*$ .

It turns out, that  $\phi$  must be measurable too. This is a direct consequence of the pipeline lemma.

**Theorem 3.2.** Given two finite-dimensional vector spaces V and W, then:

$$\mathbb{B}_{V \times W} = \mathbb{B}_V \otimes \mathbb{B}_W \tag{3.3}$$

*Proof.* Let  $e_1, e_2, \ldots, e_n$  be a basis for V with corresponding coordinate map  $\phi$ . And  $f_1, f_2, \ldots, f_m$  a basis for W with corresponding coordinate map  $\psi$ . Then  $(e_1, 0), (e_2, 0), \ldots, (e_n, 0), (0, f_1), (0, f_2), \ldots, (0, f_m)$  is a basis for  $V \times W$ . The corresponding coordinate map is:

$$\phi \times \psi : (x_1, x_2, \dots, x_{n+m}) \mapsto \left(\sum_{i=1}^n x_i e_i, \sum_{j=1}^m x_{n+j} f_j\right)$$
 (3.4)

The inverse is  $(\phi \times \psi)^{-1} = \phi^{-1} \times \psi^{-1}$ . Since  $\phi^{-1}$  is  $\mathbb{B}_V - \mathbb{B}_n$ -measurable and  $\psi^{-1}$  is  $\mathbb{B}_W - \mathbb{B}_m$ -measurable,  $\phi^{-1} \times \psi^{-1}$  must be  $\mathbb{B}_V \otimes \mathbb{B}_W - \mathbb{B}_n \otimes \mathbb{B}_n$ -measurable. But  $\mathbb{B}_n \otimes \mathbb{B}_m = \mathbb{B}_{n+m}$ . Since  $\mathbb{B}_{V \times W}$  is the smallest  $\sigma$ -algebra to make  $\phi^{-1} \times \psi^{-1}$  measurable, we must have  $\mathbb{B}_{V \times W} \subseteq \mathbb{B}_V \otimes \mathbb{B}_W$ .

On the other hand, consider the projection operators:

$$\pi_V : V \times W \to V, (v, w) \mapsto v$$
 (3.5)

$$\pi_n : \mathbb{R}^{n+m} \to \mathbb{R}^n, (x_1, \dots, x_{n+m}) \mapsto (x_1, \dots, x_n)$$
(3.6)

Now consider  $\pi_V \circ (\phi \times \psi)$ . Applied to an  $x \in \mathbb{R}^{n+m}$  we have:

$$\pi_V \circ (\phi \times \psi)(x) = \pi_v \left( \left( \sum_{i=1}^n x_i e_i, \sum_{j=1}^m x_{n+j} f_j \right) \right) = \sum_{i=1}^n x_i e_i$$
 (3.7)

But this is the same as:

$$\phi \circ \pi_1(x) = \phi((x_1, \dots x_n)) = \sum_{i=1}^n x_i e_i$$
 (3.8)

So  $\pi_V \circ (\phi \times \psi) = \phi \circ \pi_1$ . Now apply  $\phi^{-1} \times \psi^{-1}$  from the right to get:

$$\pi_V = \phi \circ \pi_1 \circ (\phi^{-1} \times \psi^{-1}) \tag{3.9}$$

Since all the three functions on the right side are measurable,  $\pi_V$  must be  $\mathbb{B}_{V\times W} - \mathbb{B}_V$ -measurable. By a similar argument the corresponding projection operator  $\pi_W: V\times W\to W$  is  $\mathbb{B}_{V\times W} - \mathbb{B}_W$ -measurable. Since  $\mathbb{B}_V\otimes \mathbb{B}_W$  is the smallest  $\sigma$ -algebra to make both  $\pi_V$  and  $\pi_W$  measurable, we must have:  $\mathbb{B}_V\otimes \mathbb{B}_W\subseteq \mathbb{B}_{V\times W}$ .

#### 3.3 Lebesgue measures on V

We now want to define a measure on the measurable space  $(V, \mathbb{B}_V)$ . If  $e_1, e_2, \ldots, e_n$  is a basis for V, we will use the associated coordinate map  $\phi$  to define a measure:

$$\lambda_V = \phi(m_n) \tag{3.10}$$

Here,  $m_n$  is the usual Lebesgue measure in n dimensions. The problem is, that this measure depends on the chosen basis! Consider another basis  $e_1^*, e_2^*, \ldots, e_n^*$  and associated coordinate map  $\phi^*$ . Then the measure is:

$$\lambda_V^* = \phi^*(m_n) = (\phi \circ \phi^{-1}) \circ \phi^*(m_n) = \phi \circ (\phi^{-1} \circ \phi^*(m_n))$$
 (3.11)

Now  $\phi^{-1} \circ \phi^*$  is an isomorphism  $\mathbb{R}^n \to \mathbb{R}^n$ , so according to section 1.1, there is a constant c such that  $(\phi^{-1} \circ \phi^*(m_n)) = cm_n$ . So:

$$\lambda_V^* = c\phi(m_n) = c\lambda_V \tag{3.12}$$

So while there are many Lebesgue measures on V they only differ from each other by a constant factor. This means that they all agree on what constitutes a null set, and on which functions are integrable. They disagree on the integral, but agree on whether it is finite or not. The also agree on whether a measure  $\mu$  has a density with respect to  $\lambda_V$  or not.