

# LDL<sup>T</sup> and Cholesky Decompositions

L22

This amounts to an LU decomposition of a symmetric, positive-definite matrix that is twice as fast to compute!

Thm: A positive-definite, symmetric matrix  $S$  can be uniquely decomposed as

$$S = LDL^T \quad \text{and} \quad S = L_1 L_1^T \quad \text{Cholesky}$$

where:

- $D$  is diagonal with positive diagonal entries
- $L$  is lower-unitriangular
- $L_1$  is lower-triangular with positive diagonal entries

See the **supplement** for a proof - it's basically Gram-Schmidt, with  $x \cdot y$  replaced by  $x^T S y$ .

NB:  $L_1$  has full column rank so  $S = L_1 L_1^T$  is necessarily positive-definite and symmetric! (L21)

NB: Let  $U = DLT$ .

(scale the rows of  $L^T$  by the diagonal entries of  $D$ )

Then  $U$  is upper-triangular with positive diagonal entries, so  $U$  is in REF, so

$$S = L(DLT) = LU \text{ is the LU decomposition.}$$

This tells us how to compute the  $LDL^T$  decomposition.

How to Compute  $S = LDL^T$ :

Let  $S$  be a symmetric matrix.

(1) Compute the  $LU$  decomposition  $S = LU$ .

→ If you have to do a row swap then **stop**:  
 $S$  is not positive-definite.

→ If the diagonal entries of  $U$  are not all positive then **stop**:  $S$  is not positive-definite.

(2) Let  $D$  = the matrix of diagonal entries of  $U$ .

(Set the off-diagonal entries to zero.) Then

$U = DLT$  (**magic!**) and  $S = LDL^T$ .

**NB:** This is the wrong procedure — it doesn't take advantage of the fact that  $S$  is symmetric. If you're more clever, you can compute  $S = LDL^T$  in  $\frac{1}{3}n^3$  time, as opposed to  $\frac{2}{3}n^3$  for  $LU$ . See the supplement if you want to know how.

**NB:** This is still an  $LU$  decomposition, so it lets you solve  $Sx = b$  in  $O(n^2)$  time.

**NB:**  $S = QDQ^T$  and  $S = LDL^T$  are both "diagonalizations" in the sense of quadratic forms — more on the HW.

Eg: Find the  $S = LDL^T$  decomposition of

$$S = \begin{pmatrix} 2 & 4 & -2 \\ 4 & 9 & -1 \\ -2 & -1 & 14 \end{pmatrix}$$

We use the 2-column method:

	$L$	$U$
	$\begin{pmatrix} & & \\ & & \\ & & \end{pmatrix}$	$\begin{pmatrix} 2 & 4 & -2 \\ 4 & 9 & -1 \\ -2 & -1 & 14 \end{pmatrix}$
$\underbrace{R_2 - 2R_1}_{R_2 \leftarrow R_2 - 2R_1}$	$\begin{pmatrix} 2 & & \\ -1 & & \\ & & \end{pmatrix}$	$\begin{pmatrix} 2 & 4 & -2 \\ 0 & 1 & 3 \\ 0 & 3 & 12 \end{pmatrix}$
$\underbrace{R_3 + R_2}_{R_3 \leftarrow R_3 + R_2}$	$\begin{pmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ -1 & 3 & 1 \end{pmatrix}$	$\begin{pmatrix} 2 & 4 & -2 \\ 0 & 1 & 3 \\ 0 & 0 & 3 \end{pmatrix}$

So  $S = LDL^T$  for

$$L = \begin{pmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ -1 & 3 & 1 \end{pmatrix} \quad D = \begin{pmatrix} 2 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 3 \end{pmatrix}$$

Check:

$$DL^T = \begin{pmatrix} 2 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 3 \end{pmatrix} \begin{pmatrix} 1 & 2 & -1 \\ 0 & 1 & 3 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 2 & 4 & -2 \\ 0 & 1 & 3 \\ 0 & 0 & 3 \end{pmatrix} \stackrel{\text{(match!)}}{=} U$$

Cholesky from  $LDL^T$ :

If  $S$  is positive definite then  $S = LDL^T$  where  $D$  is diagonal with positive diagonal entries.

$$\text{If } D = \begin{pmatrix} d_1 & & 0 \\ & \ddots & \\ 0 & & d_n \end{pmatrix} \text{ set } \sqrt{D} = \begin{pmatrix} \sqrt{d_1} & & 0 \\ & \ddots & \\ 0 & & \sqrt{d_n} \end{pmatrix}$$

Then  $\sqrt{D} \cdot \sqrt{D}^T = D$  and  $\sqrt{D}^T = \sqrt{D}$ , so

$$LDL^T = L\sqrt{D}\sqrt{D}^T L^T = (L\sqrt{D})(L\sqrt{D})^T$$

So just set

$$L_1 = L\sqrt{D} \Rightarrow S = L_1 L_1^T$$

Strong: " $S = A^T A$  is how a positive-definite symmetric matrix is put together."

$S = L_1 L_1^T$  is how you pull it apart."

Eg:  $\begin{pmatrix} 2 & 4 & -2 \\ 4 & 9 & -1 \\ -2 & -1 & 14 \end{pmatrix} = L_1 L_1^T$  for

$$L_1 = \begin{pmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ -1 & 3 & 1 \end{pmatrix} \begin{pmatrix} \sqrt{2} & 0 & 0 \\ 0 & \sqrt{5} & 0 \\ 0 & 0 & \sqrt{3} \end{pmatrix} = \begin{pmatrix} \sqrt{2} & 0 & 0 \\ 2\sqrt{2} & 1 & 0 \\ -\sqrt{2} & 3 & \sqrt{3} \end{pmatrix}$$

# Diagonalizing Quadratic Forms

In the PCA we will be interested in minimizing/maximizing the following kind of function.

**Def:** A quadratic form in  $n$  variables is a function

$$q(x_1, x_2, \dots, x_n) = \text{sum of terms of the form } a_{ij}x_i x_j.$$

Eg:  $q(x_1, x_2) = \frac{1}{2}x_1^2 + \frac{1}{2}x_2^2 - 5x_1 x_2$

Non-Eg:  $q(x_1, x_2) = x_1^2 + x_2^2 + x_1 + x_2$  is not quadratic -  
 $x_1, x_2$  are linear terms

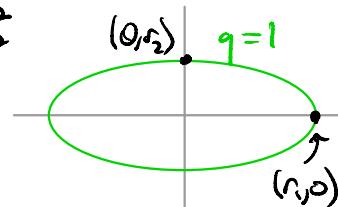
NB: Thinking of  $x = (x_1, x_2, \dots, x_n)$  as a vector in  $\mathbb{R}^n$ , for any scalar  $c$ ,

$$\begin{aligned} q(cx) &= q(cx_1, \dots, cx_n) = \sum a_{ij}(cx_i)(cx_j) \\ &= c^2 \sum a_{ij}x_i x_j = c^2 q(x) \end{aligned}$$

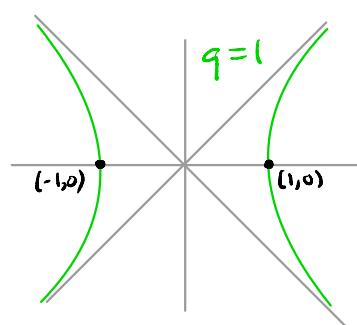
$$q(cx) = c^2 q(x)$$

Eg:  $q(x_1, x_2) = \left(\frac{x_1}{r_1}\right)^2 + \left(\frac{x_2}{r_2}\right)^2 = \frac{1}{r_1^2}x_1^2 + \frac{1}{r_2^2}x_2^2$

is a quadratic form and  $q(x_1, x_2) = 1$  defines an ellipse centered at the origin with radii  $r_1$  &  $r_2$ .



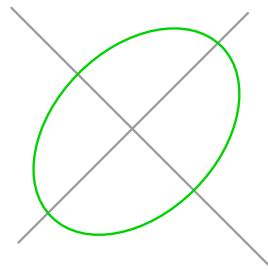
Eg:  $q(x_1, x_2) = x_1^2 - x_2^2$  is a quadratic form and  $q(x_1, x_2) = 1$  defines a hyperbola. Indeed, you can factor  $x_1^2 - x_2^2 = (x_1 - x_2)(x_1 + x_2) = 1$ , so this is  $xy = 1$  for  $x = x_1 - x_2$ ,  $y = x_1 + x_2$  (change of variables)



Eg:  $q(x_1, x_2) = \frac{5}{3}x_1^2 + \frac{5}{3}x_2^2 - x_1x_2 = 1$

also defines a conic section.

Is it an ellipse or a hyperbola?  
What are the axes & radii?



This example is a lot harder because the quadratic form had a **cross-term**:

$$q(x_1, x_2) = \frac{5}{3}x_1^2 + \frac{5}{3}x_2^2 - x_1x_2$$

The quadratic forms with no cross-terms are the easiest to understand.

Def: A quadratic form is **diagonal** if it has no cross-terms. In other words, it has the form

$$q(x_1, \dots, x_n) = \lambda_1 x_1^2 + \lambda_2 x_2^2 + \dots + \lambda_n x_n^2.$$

Eg: We can make a **linear change of variables** to eliminate the cross term in

$$q(x_1, x_2) = \frac{5}{2}x_1^2 + \frac{5}{2}x_2^2 - x_1x_2.$$

Set  $x_1 = \frac{1}{\sqrt{2}}(y_1 + y_2)$ ,  $x_2 = \frac{1}{\sqrt{2}}(y_1 - y_2)$ . Then

$$q(x_1, x_2) = q\left(\frac{1}{\sqrt{2}}(y_1 + y_2), \frac{1}{\sqrt{2}}(y_1 - y_2)\right)$$

$$= \frac{5}{2} \cdot \frac{1}{2}(y_1 + y_2)^2 + \frac{5}{2} \cdot \frac{1}{2}(y_1 - y_2)^2 - \frac{1}{2}(y_1 + y_2)(y_1 - y_2)$$

$$= \frac{5}{4}(y_1^2 + y_2^2 + 2y_1y_2) + \frac{5}{4}(y_1^2 + y_2^2 - 2y_1y_2) - \frac{1}{2}(y_1^2 - y_2^2)$$

$$= \frac{5}{2}(y_1^2 + y_2^2) - \frac{1}{2}(y_1^2 - y_2^2)$$

$$= 2y_1^2 + 3y_2^2$$

This is diagonal!

How did I know to change coordinates like this?  
Here's how to turn it into a question about symmetric matrices.

**Fact:** Every quadratic form can be written

$$q(x) = x^T S x$$

for a symmetric matrix  $S$ .

**NB:**  $x^T S x = x \cdot (Sx)$  is a scalar.

$$\text{Eg: } S = \begin{pmatrix} 1 & 2 & 3 \\ 2 & 4 & 5 \\ 3 & 5 & 6 \end{pmatrix}$$

$$\begin{aligned}
 x^T S x &= (x_1 \ x_2 \ x_3) \begin{pmatrix} 1 & 2 & 3 \\ 2 & 4 & 5 \\ 3 & 5 & 6 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} \\
 &= (x_1 \ x_2 \ x_3) \begin{pmatrix} 1x_1 + 2x_2 + 3x_3 \\ 2x_1 + 4x_2 + 5x_3 \\ 3x_1 + 5x_2 + 6x_3 \end{pmatrix} \\
 &= 1x_1^2 + 2x_1x_2 + 3x_1x_3 \\
 &\quad + 2x_2x_1 + 4x_2^2 + 5x_2x_3 \\
 &\quad + 3x_3x_1 + 5x_3x_2 + 6x_3^2 \\
 &= 1x_1^2 + 4x_2^2 + 6x_3^2 + 4x_1x_2 + 6x_1x_3 + 10x_2x_3
 \end{aligned}$$

NB: The (1,2) and (2,1) entries both contribute to the  $x_1x_2$  coefficient, but only the (1,1) entry contributes to the  $x_1^2$  coefficient.

How to get  $S$  from  $q$ ?

The  $x_i^2$  coefficients go on the diagonal, and half of the  $x_i x_j$  coefficient goes in each of the  $(i,j)$  and  $(j,i)$  entries:

$$q = a_{11}x_1^2 + a_{22}x_2^2 + a_{33}x_3^2 + a_{12}x_1x_2 + a_{13}x_1x_3 + a_{23}x_2x_3 \implies S = \begin{pmatrix} a_{11} & a_{12}/2 & a_{13}/2 \\ a_{12}/2 & a_{22} & a_{23}/2 \\ a_{13}/2 & a_{23}/2 & a_{33} \end{pmatrix}$$

NB:  $q$  is diagonal  $\Leftrightarrow S$  is diagonal: the  $a_{ij}$ 's are the coefficients of the cross terms. So:

$$x^T \begin{pmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \ddots & & 0 \\ \vdots & & \ddots & 0 \\ 0 & 0 & \cdots & \lambda_n \end{pmatrix} x = \lambda_1 x_1^2 + \lambda_2 x_2^2 + \cdots + \lambda_n x_n^2$$

Key Idea: let's orthogonally diagonalize  $S$ :

$$S = QDQ^T \quad D = \begin{pmatrix} \lambda_1 & 0 & \cdots & 0 \\ 0 & \ddots & & 0 \\ \vdots & & \ddots & 0 \\ 0 & 0 & \cdots & \lambda_n \end{pmatrix}$$

Let  $x = Qy$ : this is a linear change of variables.

Then

$$\begin{aligned} q(x) &= x^T S x = x^T Q D Q^T x \stackrel{x=Qy}{=} (Qy)^T Q D Q^T (Qy) \\ &= y^T Q^T Q D Q^T Q y \stackrel{Q^T Q = I_n}{=} y^T D y \\ &= \lambda_1 y_1^2 + \lambda_2 y_2^2 + \cdots + \lambda_n y_n^2. \end{aligned}$$

This is diagonal!

How to Diagonalize a Quadratic Form  $q$ :

(1) Write  $q(x) = x^T S x$  for a symmetric matrix  $S$

(hard part) (2) Orthogonally diagonalize  $S = QDQ^T$

(3) Change variables  $x = Qy$ .

Then  $q = \lambda_1 y_1^2 + \lambda_2 y_2^2 + \cdots + \lambda_n y_n^2$ , where

$\lambda_1, \lambda_2, \dots, \lambda_n$  are the eigenvalues of  $S$ .

$$\text{Eg: } q(x_1, x_2) = \frac{5}{2}x_1^2 + \frac{5}{3}x_2^2 - x_1x_2$$

$$(1) q(x) = x^T S x \text{ for } S = \begin{pmatrix} 5/2 & -1/2 \\ -1/2 & 5/3 \end{pmatrix}$$

$$(2) S = Q D Q^T \text{ for}$$

$$Q = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix} \quad D = \begin{pmatrix} 2 & 0 \\ 0 & 3 \end{pmatrix}$$

$$(3) x = Qy: \text{ this means}$$

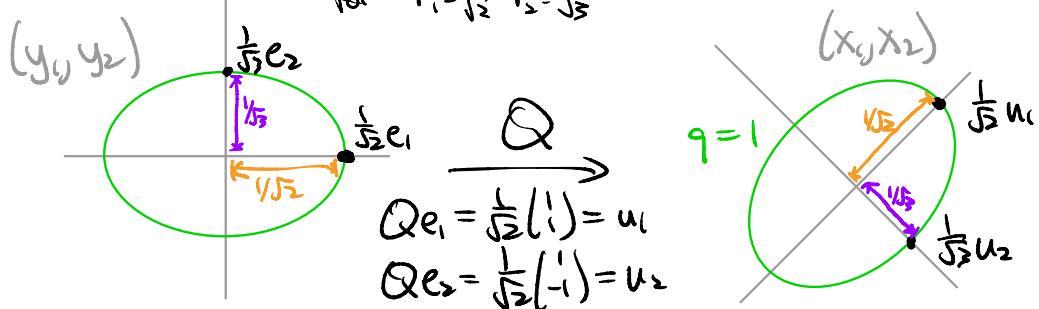
$$\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = Q \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} = \begin{pmatrix} \frac{1}{\sqrt{2}}(y_1 + y_2) \\ \frac{1}{\sqrt{2}}(y_1 - y_2) \end{pmatrix}$$

$$\Rightarrow q = 2y_1^2 + 3y_2^2$$

These explains where the formulas before came from! It also tells us how to draw the ellipse. This is easy in the  $(y_1, y_2)$ -coordinates:

$$q = 2y_1^2 + 3y_2^2 = \left(\frac{y_1}{r_1}\right)^2 + \left(\frac{y_2}{r_2}\right)^2$$

$$\text{for } r_1 = \frac{1}{\sqrt{2}}, r_2 = \frac{1}{\sqrt{3}}$$



So the axes are the lines thru  $u_1$  &  $u_2$  and the radii are  $\frac{1}{\sqrt{2}}$  and  $\frac{1}{\sqrt{3}}$ . (NB  $\|u_1\| = 1 = \|u_2\|$ )

Diagonalization also tells us if  $q(\mathbf{x})=1$  is an ellipse or a hyperbola:

- it's an **ellipse** if both eigenvalues are **positive**:

$$\lambda_1 y_1^2 + \lambda_2 y_2^2 = 1 \quad \lambda_1, \lambda_2 > 0$$

Since  $\lambda_1, \lambda_2$  are the eigenvalues of  $S$ , this means  $S$  is **positive-definite**

- it's a **hyperbola** if  $\lambda_1 > 0$  and  $\lambda_2 < 0$  or vice-versa: this means  $S$  is **indefinite**.

**Def:** A quadratic form  $q$  is **positive-definite** if  $q(\mathbf{x}) > 0$  for all  $\mathbf{x} \neq 0$ .

If  $q(\mathbf{x}) = \mathbf{x}^T S \mathbf{x}$  then  $q$  is positive-definite  $\Leftrightarrow S$  is positive-definite by the positive-energy criterion.

In this case,  $q=1$  defines an **ellipsoid** ("egg"), and orthogonally diagonalizing  $S$  computes its axes & radii.



## How to Put an Ellipsoid in Standard Form:

Let  $q$  be a positive-definite quadratic form, so  $q(x)=1$  defines an ellipsoid. Write  $q(x)=x^T S x$ .

Orthogonally diagonalize  $S$ :

$$S = Q D Q^T \quad Q = \begin{pmatrix} 1 & & & \\ & \ddots & & \\ & & 1 & \\ & & & 1 \end{pmatrix} \quad D = \begin{pmatrix} \lambda_1 & & & \\ & \ddots & & \\ & & \lambda_n & \\ & & & 0 \end{pmatrix}$$

The axes go thru  $u_1, \dots, u_n$ , and the radii are  $\sqrt{\lambda_1}, \dots, \sqrt{\lambda_n}$ .

Eg: Let's diagonalize  $q(x) = \frac{1}{5}(9x_1^2 + 6x_2^2 - 4x_1x_2)$

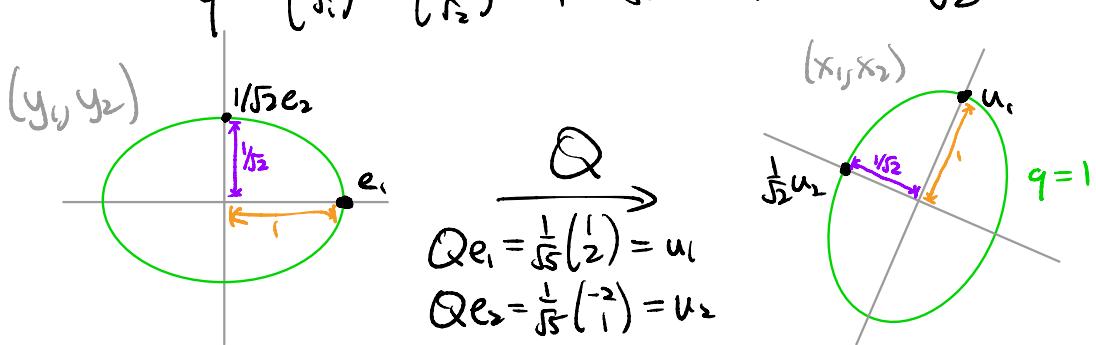
(1)  $q(x) = x^T S x$  for  $S = \frac{1}{5} \begin{pmatrix} 9 & -4 \\ -4 & 6 \end{pmatrix}$

(2)  $S = Q D Q^T$  for  $Q = \frac{1}{\sqrt{5}} \begin{pmatrix} 1 & -2 \\ 2 & 1 \end{pmatrix} \quad D = \begin{pmatrix} 1 & 0 \\ 0 & 2 \end{pmatrix}$

(3)  $x = Qy \rightsquigarrow q = y_1^2 + 2y_2^2$

This means  $q=1$  is an ellipse:

$$q = \left(\frac{y_1}{r_1}\right)^2 + \left(\frac{y_2}{r_2}\right)^2 = 1 \quad \text{for } r_1=1 \quad r_2=\frac{1}{\sqrt{2}}$$



So the axes go thru  $u_1$  &  $u_2$ , with radii 1 &  $1/\sqrt{2}$ .