

LDLT 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

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

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

where:

- D is diagonal with positive diagonal entries
- L is lower-unittriangular
- 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 $U = DL^T$.

(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

$S = L(DL^T) = LU$ 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

$U = DL^T$ (**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
	$\begin{pmatrix} & & \\ & & \\ & & \end{pmatrix}$	$\begin{pmatrix} 2 & 4 & -2 \\ 4 & 9 & -1 \\ -2 & -1 & 14 \end{pmatrix}$
$\begin{matrix} R_2 \leftarrow R_2 - 2R_1 \\ R_3 \leftarrow R_3 + R_1 \end{matrix}$	$\begin{pmatrix} & & \\ 2 & & \\ -1 & & \end{pmatrix}$	$\begin{pmatrix} 2 & 4 & -2 \\ 0 & 1 & 3 \\ 0 & 3 & 12 \end{pmatrix}$
$R_3 \leftarrow R_3 - 3R_2$	$\begin{pmatrix} 1 & 0 & 0 \\ 2 & 1 & 0 \\ -1 & 3 & 1 \end{pmatrix}$	$\begin{pmatrix} 2 & 4 & -2 \\ 0 & 1 & 3 \\ 0 & 0 & 3 \end{pmatrix}$

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:

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


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

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

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

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

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

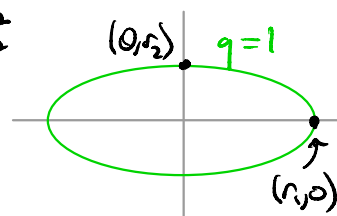
Non-Eg: $q(x_1, x_2) = x_1^2 + x_2^2 + x_1 + x_2$ is not quadratic - x_1, x_2 are linear terms

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

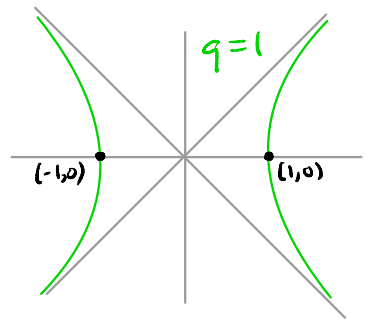
$$q(cx) = q(cx_1, \dots, cx_n) = \sum a_{ij}(cx_i)(cx_j) \\ = c^2 \sum a_{ij} x_i x_j = c^2 q(x)$$

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

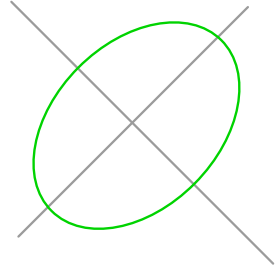
Eg: $q(x_1, x_2) = \left(\frac{x_1}{r_1}\right)^2 + \left(\frac{x_2}{r_2}\right)^2 = \frac{1}{r_1^2}x_1^2 + \frac{1}{r_2^2}x_2^2$ is a quadratic form and $q(x_1, x_2) = 1$ defines an **ellipse** centered at the origin with radii r_1 & r_2 .



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

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

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

$$\begin{aligned} q(x_1, x_2) &= q\left(\frac{1}{\sqrt{2}}(y_1 + y_2), \frac{1}{\sqrt{2}}(y_1 - y_2)\right) \\ &= \frac{5}{2} \cdot \frac{1}{2}(y_1 + y_2)^2 + \frac{5}{2} \cdot \frac{1}{2}(y_1 - y_2)^2 - \frac{1}{2}(y_1 + y_2)(y_1 - y_2) \\ &= \frac{5}{4}(y_1^2 + y_2^2 + 2y_1y_2) + \frac{5}{4}(y_1^2 + y_2^2 - 2y_1y_2) - \frac{1}{2}(y_1^2 - y_2^2) \\ &= \frac{5}{2}(y_1^2 + y_2^2) - \frac{1}{2}(y_1^2 - y_2^2) \\ &= 2y_1^2 + 3y_2^2 \end{aligned}$$

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.

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

$$x^T S x = (x_1 \ x_2 \ x_3) \begin{pmatrix} 1 & 2 & 3 \\ 2 & 4 & 5 \\ 3 & 5 & 6 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}$$

$$= (x_1 \ x_2 \ x_3) \begin{pmatrix} 1x_1 + 2x_2 + 3x_3 \\ 2x_1 + 4x_2 + 5x_3 \\ 3x_1 + 5x_2 + 6x_3 \end{pmatrix}$$

$$= 1x_1^2 + 2x_1x_2 + 3x_1x_3$$

$$+ 2x_2x_1 + 4x_2^2 + 5x_2x_3$$

$$+ 3x_3x_1 + 5x_3x_2 + 6x_3^2$$

$$= 1x_1^2 + 4x_2^2 + 6x_3^2 + 4x_1x_2 + 6x_1x_3 + 10x_2x_3$$

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_ix_j coefficient goes in each of the (i,j) and (j,i) entries:

$$q = a_{11}x_1^2 + a_{22}x_2^2 + a_{33}x_3^2$$

$$+ a_{12}x_1x_2 + a_{13}x_1x_3 + a_{23}x_2x_3$$

$$\leadsto S = \begin{pmatrix} a_{11} & a_{12}/2 & a_{13}/2 \\ a_{12}/2 & a_{22} & a_{23}/2 \\ a_{13}/2 & a_{23}/2 & a_{33} \end{pmatrix}$$

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

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

Key Idea: Let's orthogonally diagonalize S :

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

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

Then

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

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 $q = \lambda_1 y_1^2 + \lambda_2 y_2^2 + \dots + \lambda_n y_n^2$, where $\lambda_1, \lambda_2, \dots, \lambda_n$ are the eigenvalues of S .

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

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

(2) $S = Q D Q^T$ 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) $x = Qy$: this means

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

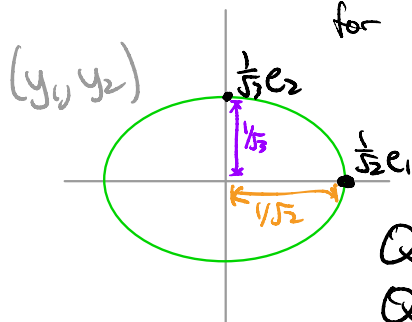
$$\rightarrow q = 2y_1^2 + 3y_2^2$$

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:

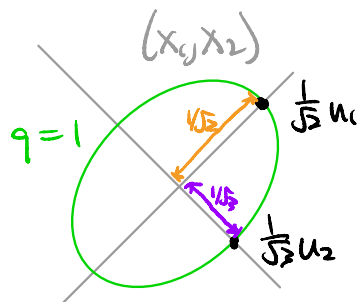
$$q = 2y_1^2 + 3y_2^2 = \left(\frac{y_1}{1/\sqrt{2}}\right)^2 + \left(\frac{y_2}{1/\sqrt{3}}\right)^2$$

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



$$Q e_1 = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix} = u_1$$

$$Q e_2 = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ -1 \end{pmatrix} = u_2$$



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 $q(x)=1$ is an ellipse or a hyperbola:

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

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

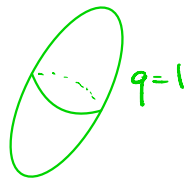
Since λ_1, λ_2 are the eigenvalues of S , this means S is **positive-definite**

- 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 $q(x) > 0$ for all $x \neq 0$.

If $q(x) = x^T S x$, 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 :

$$S = Q D Q^T \quad Q = \begin{pmatrix} | & & | \\ u_1 & \dots & u_n \\ | & & | \end{pmatrix} \quad D = \begin{pmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{pmatrix}$$

The axes go thru u_1, \dots, u_n , and the radii are $1/\sqrt{\lambda_1}, \dots, 1/\sqrt{\lambda_n}$.

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

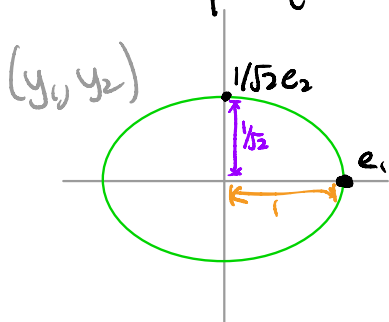
(1) $q(x) = x^T S x$ for $S = \frac{1}{5} \begin{pmatrix} 9 & -2 \\ -2 & 6 \end{pmatrix}$

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

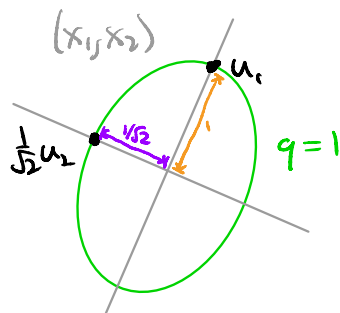
(3) $x = Q y \rightsquigarrow q = y_1^2 + 2y_2^2$

This means $q=1$ is an ellipse:

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



$$\begin{aligned} Q e_1 &= \frac{1}{\sqrt{5}} \begin{pmatrix} 1 \\ 2 \end{pmatrix} = u_1 \\ Q e_2 &= \frac{1}{\sqrt{5}} \begin{pmatrix} -2 \\ 1 \end{pmatrix} = u_2 \end{aligned}$$



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