

# Derivative of the Polar Decomposition

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We consider the derivative of the polar decomposition of a “time-varying” matrix  $A(t)$ :

$$A(t) = R(t)S(t) \tag{1}$$

where  $R$  is the closest rotation matrix to  $A$  (in Frobenius norm) and  $S$  is a symmetric matrix capturing the stretching performed by  $A$ . Notice that if  $A$  inverts space (if  $\det(A) < 0$ ), then this decomposition is actually *not* the polar decomposition but rather the closely related decomposition where the smallest eigenvalue of  $S$  is negated.

We begin by differentiating both sides of (1), denoting quantities’ instantaneous rates of change with a dot (e.g.,  $\dot{A} := \frac{d}{dt} \Big|_{t=0} A(t)$ ):

$$\dot{A} = \dot{R}S + R\dot{S} \iff R^T \dot{A} = R^T \dot{R}S + \dot{S}. \tag{2}$$

We observe that because  $S(t)$  is symmetric for all  $t$ ,  $\dot{S}$  must be a symmetric matrix. Furthermore, because  $R^T R = I$ , we find by differentiating both sides that  $R^T \dot{R}$  is a skew symmetric matrix:  $\dot{R}^T R + R^T \dot{R} = 0$ . We can use these symmetric and skew symmetric properties to isolate and solve for  $\dot{R}$ ; computing the skew symmetric part of both sides kills off the symmetric  $\dot{S}$  term:

$$\begin{aligned} R^T \dot{A} - (R^T \dot{A})^T &= R^T \dot{R}S - (R^T \dot{R}S)^T + \cancel{\dot{S} - \dot{S}^T} \\ \underbrace{R^T \dot{A} - \dot{A}^T R^T}_C &= \underbrace{R^T \dot{R}S}_M - S (R^T \dot{R})^T = MS - SM^T, \end{aligned}$$

which is a *Sylvester equation*  $C = MS + SM$  for skew symmetric matrix  $M := R^T \dot{R}$ . We solve this equation by using  $S$ ’s eigen decomposition  $S = Q\Lambda Q^T$ :

$$C = MQ\Lambda Q^T + Q\Lambda Q^T M \iff Q^T C Q = Q^T M Q \Lambda + \Lambda Q^T M Q.$$

It becomes clear that this equation is now easy to solve when we inspect its components:

$$[Q^T C Q]_{ij} = [Q^T M Q]_{ij} \lambda_j + \lambda_i [Q^T M Q]_{ij} = (\lambda_i + \lambda_j) [Q^T M Q]_{ij}. \tag{3}$$

(In this equation  $i, j$  are free indices and summation is *not* implied.) In other words, we simply divide the  $ij^{\text{th}}$  component of  $Q^T C Q$  by  $\lambda_i + \lambda_j$  to find  $Q^T M Q$ . From here we are essentially done: we can just compute  $\dot{R} = RM = RQ(Q^T M Q)Q^T$ .

We can express the division operation from (3) in terms of standard linear algebra operations by converting matrix equation (3) into a vector equation. First, we notice that both  $Q^T C Q$  and  $Q^T M Q$  are skew symmetric matrices because both  $C$  and  $M$  are. Any skew symmetric matrix  $B$  has the form

$$B = \begin{bmatrix} 0 & -b_3 & b_2 \\ b_3 & 0 & -b_1 \\ -b_2 & b_1 & 0 \end{bmatrix},$$

where we chose the signs so that  $B$  is actually the cross product matrix for vector  $\mathbf{b} = (b_1, b_2, b_3)^T$ . In other words, we have the property that  $B\mathbf{v} = \mathbf{b} \times \mathbf{v}$  for all vectors  $\mathbf{v}$ .

We can introduce the linear operators  $\mathbf{sk}(\cdot)$  and  $\mathbf{sk}^{-1}(\cdot)$  to convert between a skew symmetric matrix and its corresponding cross-product vector:

$$\mathbf{sk} \left( \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} \right) = \begin{bmatrix} 0 & -b_3 & b_2 \\ b_3 & 0 & -b_1 \\ -b_2 & b_1 & 0 \end{bmatrix}, \quad \mathbf{sk}^{-1} \left( \begin{bmatrix} 0 & -b_3 & b_2 \\ b_3 & 0 & -b_1 \\ -b_2 & b_1 & 0 \end{bmatrix} \right) = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}.$$

Applying  $\mathbf{sk}^{-1}$  to both sides of (3), we obtain:

$$\mathbf{sk}^{-1}(Q^T C Q) = \begin{pmatrix} \lambda_2 + \lambda_3 & 0 & 0 \\ 0 & \lambda_1 + \lambda_3 & 0 \\ 0 & 0 & \lambda_1 + \lambda_2 \end{pmatrix} \mathbf{sk}^{-1}(Q^T M Q) = (\text{tr}(\Lambda)I - \Lambda) \mathbf{sk}^{-1}(Q^T M Q).$$

Next, we apply the transformation formula  $\mathbf{sk}^{-1}(Q^T B Q) = Q^T \mathbf{sk}^{-1}(B)$  (rotating, taking a cross product with  $\mathbf{b}$ , and then rotating back is the same thing as rotating  $\mathbf{b}$  back and taking a cross product):

$$\begin{aligned} Q^T \mathbf{sk}^{-1}(C) &= (\text{tr}(\Lambda)I - \Lambda) Q^T \mathbf{sk}^{-1}(M) \iff \\ \mathbf{sk}^{-1}(C) &= Q(\text{tr}(\Lambda)I - \Lambda) Q^T \mathbf{sk}^{-1}(M) = (\text{tr}(S)I - S) \mathbf{sk}^{-1}(M) \iff \\ M &= \mathbf{sk} \left( (\text{tr}(S)I - S)^{-1} \mathbf{sk}^{-1}(C) \right). \end{aligned}$$

Finally, we plug in the expressions for  $C$  and  $M$  and solve for  $\dot{R}$ :

$$\begin{aligned} R^T \dot{R} &= (\text{tr}(S)I - S)^{-1} \mathbf{sk}^{-1}(R^T \dot{A} - \dot{A}^T R) \iff \\ \dot{R} &= R \mathbf{sk} \left( 2(\text{tr}(S)I - S)^{-1} \mathbf{sk}^{-1}(R^T \dot{A}) \right), \end{aligned}$$

where we've extended the definition of  $\mathbf{sk}^{-1}(B)$  to non skew-symmetric matrices  $B$  by having it operate on the skew symmetric part  $\frac{1}{2}(B - B^T)$ .

Now that we know  $\dot{R}$ , it's easy to solve for  $\dot{S}$  by going back to (2):

$$\dot{S} = R^T \dot{A} - R^T \dot{R} S = R^T (\dot{A} - \dot{R} S).$$