

# Solving A Rayleigh Quotient

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The goal is to solve

$$\max_w \frac{w^T S_1 w}{w^T S_2 w} \quad (1)$$

- Given the assumption that  $S_1, S_2$  are symmetric positive semi-definite (SPSD)
- First note that SPD matrices can "always" be decomposed via eigen decomposition where

$$S_i = U \Sigma U^T$$

This also implies that  $S_i$  can also be split into just 2 matrices where

$$\begin{aligned} S_i &= U \Sigma U^T \\ &= \underbrace{U \Sigma^{1/2}}_{R_i^T} \underbrace{\Sigma^{1/2} U^T}_{R_i} \end{aligned}$$

$$= R_i^T R_i \quad (2)$$

Using Eq (2) in (1), we get

$$\max_w \frac{w^T R_1^T R_1 w}{w^T R_2^T R_2 w} \quad (3)$$

We next let

$$v = R_2 w \quad (4)$$

which implies

$$R_2^{-1} v = w \quad (5)$$

Using (4), (5) in (3) we get

$$\max_w \frac{v^T (R_2^{-1})^T R_1^T R_1 (R_2^{-1}) v}{v^T v}$$

This is the same as

$$\max_w \frac{[R_1 (R_2^{-1}) v]^2}{\|v\|^2}$$

$$= \frac{v^T R_1^T R_1 v}{v^T v}$$

$$\max_w \left[ \frac{R_1(R_2^{-1})v}{\|v\|} \right]^2$$

$$\max_w \underbrace{\left[ R_1(R_2^{-1}) \frac{v}{\|v\|} \right]^2}_Q$$

Note  $\frac{v}{\|v\|}$  is a unit vector projecting onto  $R_1(R_2^{-1})$ .

To make  $Q$  as large as possible we want

$$\left[ R_1(R_2^{-1}) \right]^T = \frac{v}{\|v\|}$$

Therefore the  $v$  that maximize  $Q$  is some constant multiple  $\alpha$  with

$$\alpha \left[ R_1(R_2^{-1}) \right]^T = v \quad (6)$$

From Eq (5) we know that

$$R_2^{-1}v = w \quad (5)$$

$$R_2^T V = W \quad (5)$$

Since we know the best  $V$   
we can plug (6) into (5)

$$\alpha R_2^{-1} [R_1 (R_2^{-1})]^T = W$$

$$\alpha R_2^{-1} (R_2^{-1})^T R_1^T = W$$

$$\alpha R_2^{-1} (R_2^T)^{-1} R_1^T = W$$

$$\alpha [R_2^T R_2]^{-1} R_1^T = W$$

$$\alpha S_2^{-1} R_1^T = W^*$$

This is the optimal solution  
for Rayleigh Quotient.

In addition, note that if  
we plug the optimal  $W^*$  into  
the original objective, we get

$$\max_W \frac{W^T S_1 W}{W^T S_2 W}$$

$$\max_W \frac{W^T (S_1 - \lambda S_2) W}{W^T S_2 W}$$

$$\max_w \frac{(\alpha S_2^{-1} R_1^T)^T S_1 (\alpha S_2^{-1} R_1^T)}{(\alpha S_2^{-1} R_1^T)^T S_2 (\alpha S_2^{-1} R_1^T)}$$

$$\max_w \frac{\cancel{\alpha^2} (S_2^{-1} R_1^T)^T S_1 (S_2^{-1} R_1^T)}{\cancel{\alpha^2} (S_2^{-1} R_1^T)^T S_2 (S_2^{-1} R_1^T)}$$

Notice that  $\alpha$  can be cancelled  
so any number for  $\alpha$ , we would  
get the same objective value.

It is then easiest to just set it  
to 1, leading us to the final solution

$$S_2^{-1} R_1^T = w^*$$