

Equalizing what should be equal

—

Solving the even eigenvalue problem by
transforming an even URV form to even Schur form

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The Even Eigenvalue Problem

- ▶ special case of generalized eigenvalue problem

$$Mx = \lambda Nx$$

$M = M^T$ symmetric, $N = -N^T$ skew symmetric

- ▶ $M, N \in \mathbb{C}^{n,n}$, dense, $(\cdot)^T$ denotes the complex transpose (*-case, real case are similar, but different)
- ▶ is called: *even* eigenvalue problem, as

$$P(\lambda) := \lambda N - M = P(-\lambda)^T$$

- ▶ related to Hamiltonian eigenvalue problem \Rightarrow similar methods for this talks topic in particular: [Wat06]¹

¹David Watkins, *On the reduction of a Hamiltonian matrix to Hamiltonian Schur form*, Electron. Trans. Numer. Anal. 23, 2006

Application: The LQ optimal control problem

- ▶ dynamic system (matrices E, A, B given):

$$E\dot{x}(t) = Ax(t) + Bu(t), \quad x(0) = x_0$$

choosing $u(t)$ determines $x(t)$

- ▶ problem: chose $u(t)$ which minimizes
$$\int_0^\infty x(t)^T Qx(t) + 2x(t)^T Su(t) + u(t)^T Ru(t) dt$$

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- ▶ yields eigenvalue problem for

$$\lambda \underbrace{\begin{bmatrix} 0 & -E & 0 \\ E^T & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}}_{N=-N^T} + \underbrace{\begin{bmatrix} 0 & A & B \\ A^T & Q & S \\ B^T & S^T & R \end{bmatrix}}_{M=M^T}$$

- ▶ needed: deflating subspace for eigenvalues with negative real part

Why not use standard algorithms?

- ▶ system features spectral symmetry:

$$Mx = \lambda Nx \quad \xleftrightarrow{(\cdot)^T} \quad x^T M = (-\lambda)x^T N$$

i.e., also $-\lambda$ is eigenvalue \Rightarrow pairs $\pm\lambda$

- ▶ structure is important for applications (positive/negative real part)
- ▶ general algorithms (like the QZ algorithm) destroy this eigenvalue pairing due to rounding errors

Special case: skew triangular

- ▶ If M, N are skew triangular, i.e., $m_{ij} = 0$ whenever $i + j \leq n$,

$$M = \begin{array}{|c} \triangle \\ \hline \end{array}, \quad N = \begin{array}{|c} \triangle \\ \hline \end{array},$$

- ▶ then

$$Me_1 = m_{n1}e_n, \quad Ne_1 = n_{n1}e_n$$

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- ▶ and more general

$[e_1, e_2, \dots, e_k]$ span right deflating subspace,

$[e_{n-k+1}, \dots, e_n]$ span left deflating subspace,

$\frac{m_{n-i+1,i}}{n_{n-i+1,i}}$ are eigenvalues ($i = 1, \dots, k$).

- ▶ So, our problem is already solved.
- ▶ What if M, N are not skew triangular?

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- ▶ What we can compute: unitary U, V such that

$$U^T N U = \triangleleft, \quad \underbrace{U^T M V}_R = \triangleleft, \quad V^T N V = \triangleleft$$

called *even URV form* (as $M = \bar{U} R V^*$); algorithm: [Sch07]²; provides eigenvalues, eigenvectors, not subspaces

²C. Schröder, *URV decomposition based structured methods for palindromic and even eigenvalue problems*, MATHEON Preprint 375, 2007

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- ▶ note reduces to even Schur form if $U = V$
- ▶ Idea: transform a URV form to Schur form \Rightarrow
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Today: first step

- ▶ Given an even URV form

$$U^T N U = T = \triangleleft, \quad U^T M V = R = \triangleleft, \quad V^T N V = P = \triangleleft$$

modify U, V to $\tilde{U} = U\Delta_U, \tilde{V} = V\Delta_V$

$$\tilde{U}^T N \tilde{U} = \tilde{T} \quad , \quad \tilde{U}^T M \tilde{V} = \tilde{R} \quad , \quad \tilde{V}^T N \tilde{V} = \tilde{P} \quad ,$$

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such that 1) the first and last columns of \tilde{U} and \tilde{V} coincide

$$\tilde{u}_1 = \tilde{v}_1, \quad \tilde{u}_n = \tilde{v}_n,$$

2) $\tilde{T}, \tilde{R}, \tilde{P}$ are still skew triangular.

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- ▶ Given an even URV form

$$U^T N U = T = \triangleleft, \quad U^T M V = R = \triangleleft, \quad V^T N V = P = \triangleleft$$

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such that 1) the first and last columns of \tilde{U} and \tilde{V} coincide

$$\begin{aligned} \tilde{u}_1 &= \tilde{v}_1, & \tilde{u}_n &= \tilde{v}_n, \\ 2) \quad \tilde{T}, \tilde{R}, \tilde{P} & \text{ are still skew triangular.} \end{aligned}$$

- ▶ remaining columns can be treated recursively
- ▶ Lets concentrate on first goal.

Goal 1: Equalizing first/last column of U, V

Note: $\tilde{u}_1 = \tilde{v}_1$ must be eigenvector, $\tilde{u}_n = \tilde{v}_n$ must be image vector.

Procedure: obtain eigen/imagevector x, y ,

then chose Δ_U, Δ_V such that $\tilde{u}_1 = \tilde{v}_1 = c_1 \cdot x$, $\tilde{u}_n = \tilde{v}_n = c_2 \cdot y$

Goal 1 achieved,

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Procedure: obtain eigen/imagevector x, y , from URV form

$$M[u_1, v_1] = [\bar{u}_n, \bar{v}_n] \begin{bmatrix} 0 & r_{n1} \\ r_{1n} & 0 \end{bmatrix}$$
$$N[u_1, v_1] = [\bar{u}_n, \bar{v}_n] \begin{bmatrix} t_{n1} & 0 \\ 0 & \rho_{n1} \end{bmatrix}$$

$\Rightarrow \text{span}(u_1, v_1)$ right deflating subspace

$\Rightarrow \text{span}(\bar{u}_n, \bar{v}_n)$ left deflating subspace of (M, N)

\leadsto contain x, y with $Mx = \alpha y$, $Nx = \beta y$.

then chose Δ_U, Δ_V such that $\tilde{u}_1 = \tilde{v}_1 = c_1 \cdot x$, $\tilde{u}_n = \tilde{v}_n = c_2 \cdot y$

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this implies that

$$\Delta_U e_1 = x_u := U^* x, \quad \Delta_U e_n = y_u := U^T y$$

\Rightarrow choose Δ_U as series of Givens rotations that transform x_u to e_1 and y_u to e_n , (same for Δ_V)

Goal 1 achieved,

Goal 1: Equalizing first/last column of U, V

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$$N[u_1, v_1] = [\bar{u}_n, \bar{v}_n] \begin{bmatrix} t_{n1} & 0 \\ 0 & p_{n1} \end{bmatrix}$$

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Goal 1 achieved, **what about goal 2?**

Goal 2: Invariance of URV form

To understand, why R , T , P stay skew triangular, the following relation is essential

$$\begin{aligned}Nx &= \beta y \\ U^T N U U^* x &= U^T y \\ Tx_u &= y_u \\ \Delta_U^T T \Delta_U \Delta_U^* x_u &= \Delta_U^T y_u\end{aligned}\tag{1}$$

So, (1) stays valid under an update of form

$$\begin{aligned}x_u &\leftarrow \Delta_U^* x_u \\ y_u &\leftarrow \Delta_U^T y_u \\ T &\leftarrow \Delta_U^T T \Delta_U\end{aligned}$$

for any unitary Δ_U .

A 6-by-6 example

Given $M, N \in \mathbb{C}^{6,6}$

we compute URV form, and x_u, x_v, y_u, y_v

$T, R, P, [x_u, y_u], [x_v, y_v] =$

$$\begin{bmatrix} & & & & & \mathbf{x} \\ & & & & & \mathbf{x} \mathbf{x} \\ & & & & \mathbf{x} \mathbf{x} \mathbf{x} \\ & & \mathbf{x} \mathbf{0} \mathbf{x} \mathbf{x} \\ \mathbf{x} \mathbf{x} \mathbf{x} \mathbf{0} \mathbf{x} \\ \mathbf{x} \mathbf{x} \mathbf{x} \mathbf{x} \mathbf{x} \mathbf{0} \end{bmatrix}, \begin{bmatrix} & & & & & \mathbf{x} \\ & & & & & \mathbf{x} \mathbf{x} \\ & & & & \mathbf{x} \mathbf{x} \mathbf{x} \\ & & \mathbf{x} \mathbf{x} \mathbf{x} \mathbf{x} \\ \mathbf{x} \mathbf{x} \mathbf{x} \mathbf{x} \mathbf{x} \\ \mathbf{x} \mathbf{x} \mathbf{x} \mathbf{x} \mathbf{x} \mathbf{x} \end{bmatrix}, \begin{bmatrix} & & & & & \mathbf{x} \\ & & & & & \mathbf{x} \mathbf{x} \\ & & & & \mathbf{x} \mathbf{x} \mathbf{x} \\ & & \mathbf{x} \mathbf{0} \mathbf{x} \mathbf{x} \\ \mathbf{x} \mathbf{x} \mathbf{x} \mathbf{0} \mathbf{x} \\ \mathbf{x} \mathbf{x} \mathbf{x} \mathbf{x} \mathbf{x} \mathbf{0} \end{bmatrix}, \begin{bmatrix} \mathbf{x} \mathbf{x} \\ \mathbf{x} \mathbf{x} \\ \mathbf{x} \mathbf{x} \\ \mathbf{x} \mathbf{x} \\ \mathbf{x} \mathbf{x} \\ \mathbf{x} \mathbf{x} \end{bmatrix}, \begin{bmatrix} \mathbf{x} \mathbf{x} \\ \mathbf{x} \mathbf{x} \\ \mathbf{x} \mathbf{x} \\ \mathbf{x} \mathbf{x} \\ \mathbf{x} \mathbf{x} \\ \mathbf{x} \mathbf{x} \end{bmatrix}.$$

assumption 1: M, N nonsingular \Rightarrow skew diagonal entries of T, R, P nonzero, α, β non-zero

assumption 2: last entries of x_u, x_v and first entries of y_u, y_v are nonzero

goal: $x_u, x_v \rightarrow e_1$ $y_u, y_v \rightarrow e_6$.

A 6-by-6 example

$T, R, P, [x_u, y_u], [x_v, y_v] =$

$$\begin{bmatrix} & & & & & x \\ & & & & x & x \\ & & x & x & x & \\ & x & 0 & x & x & \\ x & x & x & 0 & x & \\ x & x & x & x & x & 0 \end{bmatrix}, \begin{bmatrix} & & & & & x \\ & & & & x & x \\ & & x & x & x & \\ & x & x & x & x & \\ x & x & x & x & x & \\ x & x & x & x & x & x \end{bmatrix}, \begin{bmatrix} & & & & & x \\ & & & & x & x \\ & & x & x & x & \\ & x & 0 & x & x & \\ x & x & x & 0 & x & \\ x & x & x & x & x & 0 \end{bmatrix}, \begin{bmatrix} x & x \\ x & x \\ x & x \\ x & x \\ x & x \\ x & x \end{bmatrix}, \begin{bmatrix} x & x \\ x & x \\ x & x \\ x & x \\ x & x \\ x & x \end{bmatrix}.$$

eliminate by Givens rotations: last entries of x_u, x_v ,
first entries of y_u, y_v

A 6-by-6 example

$T, R, P, [x_u, y_u], [x_v, y_v] =$

$$\begin{bmatrix} & & & & ? & y \\ & & & & y & y \\ & & & x & y & y \\ & & x & 0 & y & y \\ ? & y & y & y & 0 & y \\ y & y & y & y & y & 0 \end{bmatrix}, \begin{bmatrix} & & & & ? & y \\ & & & & y & y \\ & & & x & y & y \\ & & x & x & y & y \\ ? & y & y & y & y & y \\ y & y & y & y & y & y \end{bmatrix}, \begin{bmatrix} & & & & ? & y \\ & & & & y & y \\ & & & x & y & y \\ & & x & 0 & y & y \\ ? & y & y & y & 0 & y \\ y & y & y & y & y & 0 \end{bmatrix}, \begin{bmatrix} y & \mathbf{0} \\ y & y \\ x & x \\ x & x \\ y & y \\ \mathbf{0} & y \end{bmatrix}, \begin{bmatrix} y & \mathbf{0} \\ y & y \\ x & x \\ x & x \\ y & y \\ \mathbf{0} & y \end{bmatrix}$$

fill-in at (1,5) and (5,1) in R, T, P !!!!

A 6-by-6 example

$$T, R, P, [x_u, y_u], [x_v, y_v] =$$

$$\begin{bmatrix} & & & & ? & y \\ & & & & y & y \\ & & x & y & y & \\ & x & 0 & y & y & \\ ? & y & y & y & 0 & y \\ y & y & y & y & y & 0 \end{bmatrix}, \begin{bmatrix} & & & & ? & y \\ & & & & y & y \\ & & x & y & y & \\ & x & x & y & y & \\ ? & y & y & y & y & y \\ y & y & y & y & y & y \end{bmatrix}, \begin{bmatrix} & & & & ? & y \\ & & & & y & y \\ & & x & y & y & \\ & x & 0 & y & y & \\ ? & y & y & y & 0 & y \\ y & y & y & y & y & 0 \end{bmatrix}, \begin{bmatrix} y & \mathbf{0} \\ y & y \\ x & x \\ x & x \\ y & y \\ \mathbf{0} & y \end{bmatrix}, \begin{bmatrix} y & \mathbf{0} \\ y & y \\ x & x \\ x & x \\ y & y \\ \mathbf{0} & y \end{bmatrix}$$

fill-in at (1,5) and (5,1) in R , T , P !!!! really?

remember, $Tx_u = \beta y_u$ still holds.

first row $\rightarrow t_{1,5} \cdot x_{u,5} = \beta y_{u,1} = 0$, so $t_{1,5} = 0 = -t_{5,1}$.

Similar for other question marks

A 6-by-6 example

$$T, R, P, [x_u, y_u], [x_v, y_v] =$$

$$\begin{bmatrix} & & & & ? & y \\ & & & & y & y \\ & & x & y & y & \\ & x & 0 & y & y & \\ ? & y & y & y & 0 & y \\ y & y & y & y & y & 0 \end{bmatrix}, \begin{bmatrix} & & & & ? & y \\ & & & & y & y \\ & & x & y & y & \\ & x & x & y & y & \\ ? & y & y & y & y & y \\ y & y & y & y & y & y \end{bmatrix}, \begin{bmatrix} & & & & ? & y \\ & & & & y & y \\ & & x & y & y & \\ & x & 0 & y & y & \\ ? & y & y & y & 0 & y \\ y & y & y & y & y & 0 \end{bmatrix}, \begin{bmatrix} y & \mathbf{0} \\ y & y \\ x & x \\ x & x \\ y & y \\ \mathbf{0} & y \end{bmatrix}, \begin{bmatrix} y & \mathbf{0} \\ y & y \\ x & x \\ x & x \\ y & y \\ \mathbf{0} & y \end{bmatrix}$$

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Similar for other question marks

next, eliminate by Givens rotations: second last entries of x_u, x_v ,
second entries of y_u, y_v

A 6-by-6 example

$$T, R, P, [x_u, y_u], [x_v, y_v] =$$

$$\begin{bmatrix} & & & & & x \\ & ? & y & y & & \\ & & y & y & y & \\ ? & y & 0 & y & y & \\ y & y & y & 0 & y & \\ x & y & y & y & y & 0 \end{bmatrix}, \begin{bmatrix} & & & & & x \\ & ? & y & y & & \\ & & y & y & y & \\ ? & y & y & y & y & \\ y & y & y & y & y & \\ x & y & y & y & y & x \end{bmatrix}, \begin{bmatrix} & & & & & x \\ & ? & y & y & & \\ & & y & y & y & \\ ? & y & 0 & y & y & \\ y & y & y & 0 & y & \\ x & y & y & y & y & 0 \end{bmatrix}, \begin{bmatrix} x & 0 \\ y & \mathbf{0} \\ y & y \\ y & y \\ \mathbf{0} & y \\ 0 & x \end{bmatrix}, \begin{bmatrix} x & 0 \\ y & \mathbf{0} \\ y & y \\ y & y \\ \mathbf{0} & y \\ 0 & x \end{bmatrix}$$

Again, fill-in in T, R, P .

A 6-by-6 example

$$T, R, P, [x_u, y_u], [x_v, y_v] =$$

$$\begin{bmatrix} & & & & & x \\ & ? & y & y & & \\ & & y & y & y & \\ ? & y & 0 & y & y & \\ y & y & y & 0 & y & \\ x & y & y & y & y & 0 \end{bmatrix}, \begin{bmatrix} & & & & & x \\ & ? & y & y & & \\ & & y & y & y & \\ ? & y & y & y & y & \\ y & y & y & y & y & \\ x & y & y & y & y & x \end{bmatrix}, \begin{bmatrix} & & & & & x \\ & ? & y & y & & \\ & & y & y & y & \\ ? & y & 0 & y & y & \\ y & y & y & 0 & y & \\ x & y & y & y & y & 0 \end{bmatrix}, \begin{bmatrix} x & 0 \\ y & \mathbf{0} \\ y & y \\ y & y \\ \mathbf{0} & y \\ 0 & x \end{bmatrix}, \begin{bmatrix} x & 0 \\ y & \mathbf{0} \\ y & y \\ y & y \\ \mathbf{0} & y \\ 0 & x \end{bmatrix}$$

Again, fill-in in T, R, P .

again, fake!!!

second row of $Tx_u = y_u$:

$$T_{2,4}x_{u,4} = 0$$

A 6-by-6 example

$$T, R, P, [x_u, y_u], [x_v, y_v] =$$

$$\begin{bmatrix} & & & & x \\ & ? & y & y \\ & y & y & y \\ ? & y & 0 & y & y \\ y & y & y & 0 & y \\ x & y & y & y & y & 0 \end{bmatrix}, \begin{bmatrix} & & & & x \\ & ? & y & y \\ & y & y & y \\ ? & y & y & y & y \\ y & y & y & y & y \\ x & y & y & y & y & x \end{bmatrix}, \begin{bmatrix} & & & & x \\ & ? & y & y \\ & y & y & y \\ ? & y & 0 & y & y \\ y & y & y & 0 & y \\ x & y & y & y & y & 0 \end{bmatrix}, \begin{bmatrix} x & 0 \\ y & \mathbf{0} \\ y & y \\ y & y \\ \mathbf{0} & y \\ 0 & x \end{bmatrix}, \begin{bmatrix} x & 0 \\ y & \mathbf{0} \\ y & y \\ y & y \\ \mathbf{0} & y \\ 0 & x \end{bmatrix}$$

Again, fill-in in T, R, P .

again, fake!!!

second row of $Tx_u = y_u$:

$$T_{2,4}x_{u,4} = 0$$

next, eliminate by Givens rotations: third entries of y_u, y_v

A 6-by-6 example

$T, R, P, [x_u, y_u], [x_v, y_v] =$

$$\begin{bmatrix}
 & & & & & x \\
 & & & & x & x \\
 0 & y & y & y & & \\
 y & 0 & y & y & & \\
 x & y & y & 0 & x & \\
 x & x & y & y & x & 0
 \end{bmatrix},
 \begin{bmatrix}
 & & & & & x \\
 & & & & x & x \\
 ? & y & y & y & & \\
 y & y & y & y & & \\
 x & y & y & x & x & \\
 x & x & y & y & x & x
 \end{bmatrix},
 \begin{bmatrix}
 & & & & & x \\
 & & & & x & x \\
 0 & y & y & y & & \\
 y & 0 & y & y & & \\
 x & y & y & 0 & x & \\
 x & x & y & y & x & 0
 \end{bmatrix},
 \begin{bmatrix}
 x & 0 \\
 x & 0 \\
 y & \mathbf{0} \\
 ? & y \\
 0 & x \\
 0 & x
 \end{bmatrix},
 \begin{bmatrix}
 x & 0 \\
 x & 0 \\
 y & \mathbf{0} \\
 ? & y \\
 0 & x \\
 0 & x
 \end{bmatrix}$$

- ▶ fake non-zeros in x_u, x_v , because $x_u^T y_u = 0$
- ▶ no fill-in in T, P , because skew symmetry
- ▶ fake fill-in in R

A 6-by-6 example

$T, R, P, [x_u, y_u], [x_v, y_v] =$

$$\begin{bmatrix}
 & & & & & x \\
 & & & & x & x \\
 0 & y & y & y & & \\
 y & 0 & y & y & & \\
 x & y & y & 0 & x & \\
 x & x & y & y & x & 0
 \end{bmatrix},
 \begin{bmatrix}
 & & & & & x \\
 & & & & x & x \\
 ? & y & y & y & & \\
 y & y & y & y & & \\
 x & y & y & x & x & \\
 x & x & y & y & x & x
 \end{bmatrix},
 \begin{bmatrix}
 & & & & & x \\
 & & & & x & x \\
 0 & y & y & y & & \\
 y & 0 & y & y & & \\
 x & y & y & 0 & x & \\
 x & x & y & y & x & 0
 \end{bmatrix},
 \begin{bmatrix}
 x & 0 \\
 x & 0 \\
 y & \mathbf{0} \\
 ? & y \\
 0 & x \\
 0 & x
 \end{bmatrix},
 \begin{bmatrix}
 x & 0 \\
 x & 0 \\
 y & \mathbf{0} \\
 ? & y \\
 0 & x \\
 0 & x
 \end{bmatrix}$$

- ▶ fake non-zeros in x_u, x_v , because $x_u^T y_u = 0$
- ▶ no fill-in in T, P , because skew symmetry
- ▶ fake fill-in in R

remaining steps as before

A 6-by-6 example

$T, R, P, [x_u, y_u], [x_v, y_v] =$

$$\begin{bmatrix} & & & & & x \\ & & ? & y & y & \\ & & y & y & y & \\ ? & y & 0 & y & y & \\ y & y & y & 0 & y & \\ x & y & y & y & y & 0 \end{bmatrix}, \begin{bmatrix} & & & & & x \\ & & ? & y & y & \\ & & y & y & y & \\ ? & y & y & y & y & \\ y & y & y & y & y & \\ x & y & y & y & y & x \end{bmatrix}, \begin{bmatrix} & & & & & x \\ & & ? & y & y & \\ & & y & y & y & \\ ? & y & 0 & y & y & \\ y & y & y & 0 & y & \\ x & y & y & y & y & 0 \end{bmatrix}, \begin{bmatrix} x & 0 \\ y & 0 \\ \mathbf{0} & \mathbf{0} \\ 0 & \mathbf{0} \\ 0 & y \\ 0 & x \end{bmatrix}, \begin{bmatrix} x & 0 \\ y & 0 \\ \mathbf{0} & \mathbf{0} \\ 0 & \mathbf{0} \\ 0 & y \\ 0 & x \end{bmatrix}$$

and finally

A 6-by-6 example

$$T, R, P, [x_u, y_u], [x_v, y_v] =$$

$$\begin{bmatrix} & & & & & & \\ & & & & & ? & y \\ & & & & & y & y \\ & & & & x & y & y \\ & & x & 0 & y & y & \\ ? & y & y & y & 0 & y & \\ y & y & y & y & y & y & 0 \end{bmatrix}, \begin{bmatrix} & & & & & & \\ & & & & & ? & y \\ & & & & & y & y \\ & & & & x & y & y \\ & & x & x & y & y & \\ ? & y & y & y & y & y & \\ y & y & y & y & y & y & \end{bmatrix}, \begin{bmatrix} & & & & & & \\ & & & & & ? & y \\ & & & & & y & y \\ & & & & x & y & y \\ & & x & 0 & y & y & \\ ? & y & y & y & 0 & y & \\ y & y & y & y & y & y & 0 \end{bmatrix}, \begin{bmatrix} y & 0 \\ \mathbf{0} & \mathbf{0} \\ 0 & 0 \\ 0 & 0 \\ 0 & \mathbf{0} \\ 0 & y \end{bmatrix}, \begin{bmatrix} y & 0 \\ \mathbf{0} & \mathbf{0} \\ 0 & 0 \\ 0 & 0 \\ 0 & \mathbf{0} \\ 0 & y \end{bmatrix}$$

As before, $? = 0$.

A 6-by-6 example

$$T, R, P, [x_u, y_u], [x_v, y_v] =$$

$$\begin{bmatrix} & & & & ? & y \\ & & & & y & y \\ & & x & y & y & \\ & x & 0 & y & y & \\ ? & y & y & y & 0 & y \\ y & y & y & y & y & 0 \end{bmatrix}, \begin{bmatrix} & & & & ? & y \\ & & & & y & y \\ & & x & y & y & \\ & x & x & y & y & \\ ? & y & y & y & y & y \\ y & y & y & y & y & y \end{bmatrix}, \begin{bmatrix} & & & & ? & y \\ & & & & y & y \\ & & x & y & y & \\ & x & 0 & y & y & \\ ? & y & y & y & 0 & y \\ y & y & y & y & y & 0 \end{bmatrix}, \begin{bmatrix} y & 0 \\ \mathbf{0} & \mathbf{0} \\ 0 & 0 \\ 0 & 0 \\ 0 & \mathbf{0} \\ 0 & y \end{bmatrix}, \begin{bmatrix} y & 0 \\ \mathbf{0} & \mathbf{0} \\ 0 & 0 \\ 0 & 0 \\ 0 & \mathbf{0} \\ 0 & y \end{bmatrix}$$

As before, $? = 0$.

At this point: $x_u, x_v = \text{const} \cdot e_1$, and $y_u, y_v = \text{const} \cdot e_6$

\Rightarrow the first columns of U and V are multiples of each other

\Rightarrow after scaling they coincide

\Rightarrow Done!

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- ▶ not from scratch, but
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Thanks for your attention. Any questions? Enjoy the dinner!