## Continuation of Sparse Eigendecompositions

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## Basic setting

Have a  $C^k$  function

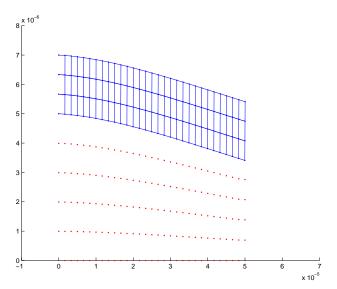
$$A: [0,1] \rightarrow \mathbb{R}^{n \times n}$$

Want to compute eigenvectors v(s) and values  $\lambda(s)$  for A(s). More generally, want an invariant subspace basis V(s).

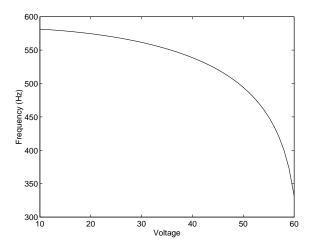
### Applications:

- 1. Resonant system design
- 2. Bifurcation analysis

## Example: Cantilever tuning



## Example: Cantilever tuning



# Example: Belousov-Zhabotinski reaction



www.pojman.com/NLCD-movies/NLCD-movies.html

### Reaction-diffusion models

$$\frac{\partial u}{\partial t} = D\nabla^2 u + F(u; s)$$

Describes many systems:

- Chemical reactions (like the B-Z reaction)
- Signals in nerves
- Ecological systems
- Phase transitions

See Chemical Oscillations, Waves, and Turbulence (Kuramoto).

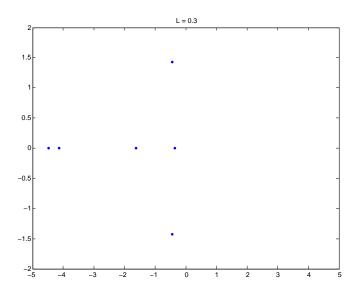
### Stability analysis

Linearize about an equilibrium branch  $u_0(s)$ :

$$\frac{\partial}{\partial t} \delta u = \left( D \nabla^2 + F_u(u_0(s); s) \right) \delta u = A(s) \delta u$$

- Stable if eigenvalues of A(s) have negative real part
- ▶ When stability changes, have a bifurcation
- ▶ Complex eigs cross imaginary axis ⇒ oscillations, a Hopf bifurcation

### Hopf bifurcation in the Brusselator



### Subspaces and stability analysis

- Diagnose stability from a small subspace (slow dynamics)
- Idea: Continue invariant subspace along with the solution
- Problem: Switching subspaces
- Problem: Missing information

## CIS algorithm

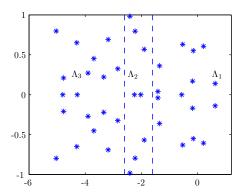
### Compute a continuous block Schur form

$$Q(s)^T A(s) Q(s) = \begin{bmatrix} T_{11}(s) & T_{12}(s) \\ 0 & T_{22}(s) \end{bmatrix}$$

#### Algorithm phases:

- Initialize
- Predict
- Correct
- Normalize
- Adapt

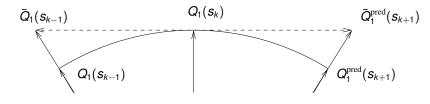
### Initialization



- Compute rightmost part of the spectrum
- Include all unstable eigenvalues + a few stable ones
- Keep eigenvalue clusters together (prevent artificially short steps)



### Prediction



Normalize to tangent plane:

$$ar{Q}_1(s) = Q_1(s) \left( Q_1(s_k)^T Q_1(s) \right)^{-1}$$

- ▶ Predict  $\bar{Q}_1(s_{k+1})$  by polynomial fitting through  $\bar{Q}_1(s_k)$ ,  $\bar{Q}_1(s_{k-1})$ , . . . .
- Suggests projection space should include computed spaces from previous few steps.



### Correction

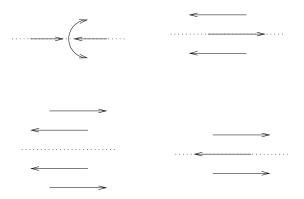
### Solve the nonlinear equations

$$AQ_1 - Q_1 T_{11} = 0$$
  
 $(Q_1^{\text{prev}})^T Q_1 - I = 0$ 

- Linearization is a bordered Sylvester equation
- ▶ Newton ≈ block RQ iteration
- ▶ Modified Newton ≈ subspace iteration
- Or extract from a Krylov subspace

Then normalize to minimize Frobenius change in  $Q_1(s)$ .

### Adaptation



### May need to adjust space if

- ▶ Real parts of continued eigenvalues overlap the rest of the spectrum (generic possibilities shown)
- Eigenvalues cross imaginary axis (bifurcation)



### Testing continuity

How to ensure proposed basis spans the right subspace?

- Check rate of Newton convergence
- Check angles between subspaces
- Check distance between eigenvalues

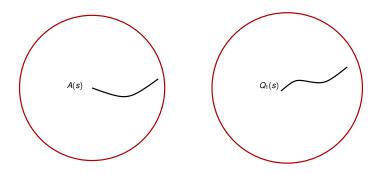
Anything less heuristic?

### Perturbation approaches



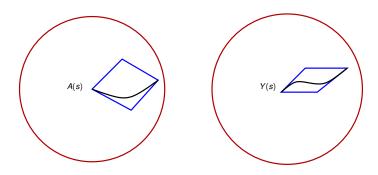
A(s) and  $Q_1(s)$  trace some paths in matrix spaces.

### Perturbation approaches



Can determine that if A(s) stays in some region,  $Q_1(s)$  is well defined and stays in some other region.

## Perturbation approaches



Can we test with smaller regions?

## Checking the subspace

Recall:

$$sep(B, C) = ||S^{-1}||^{-1}$$
, where  $S(X) = BX - XC$ 

If for  $s \in [0, h]$ ,

$$sep(A_{11},A_{22})^2 > 4\|A_{12}\|\|A_{21}\|$$

Then have a unique  $C^k$  invariant subspace associated with  $A_{11}$ .

So if 
$$A_{21}(0) = 0$$
 then  $Q_1(0) = [I; 0]$  extends to  $Q_1(s)$ 

## Checking the subspace

Have block Schur  $Q^T A(s)Q = T$  at  $s_k$  and  $s_{k+1} = s_k + h$ .

- ▶ Geodesic interpolation U(s) between  $Q(s_k)$  and  $Q(s_{k+1})$ .
- ► Similarity:  $\hat{T}(s) = U(s)^T A(s) U(s)$ .

If  $\theta_{\text{max}} = \text{largest angle between } Q(s_k) \text{ and } Q(s_{k+1}),$ 

$$\|\dot{\hat{T}}\| \leq 2\theta_{\max} \|A\| + \|\dot{A}\|.$$

Then get interpolation bounds:

- $sep(\hat{T}_{11}, \hat{T}_{22}) = O(1)$
- $\hat{T}_{12} = O(1)$
- $\hat{T}_{21} = O(h^2)$

## Checking the subspace

Check that for all  $s \in [0, h]$ ,

$$\text{sep}(\,\hat{T}_{11},\,\hat{T}_{22})^2 > 4\|\,\hat{T}_{12}\|\|\,\hat{T}_{21}\|$$

#### So test based on:

- Conditioning of subspace (spectral separation)
- ▶ Measure of non-normality ( $\|\hat{T}_{12}\|$ )
- ▶ Residual from interpolating ( $\|\hat{T}_{21}\|$ )

### Are we there yet?

Believe we can compute

$$Q(s)^T A(s) Q(s) = \begin{bmatrix} T_{11}(s) & T_{12}(s) \\ 0 & T_{22}(s) \end{bmatrix}$$

What if the space isn't rich enough ( $T_{22}$  unstable)?

Want conditions for  $T_{22}$  stable.

- Do not want another nonsymmetric eigenproblem
- Only need sufficient conditions

Use the fact that we expect rapid decay in most modes.

# Stability of $T_{22}$

Recall: if  $\dot{u} = T_{22}u$  then

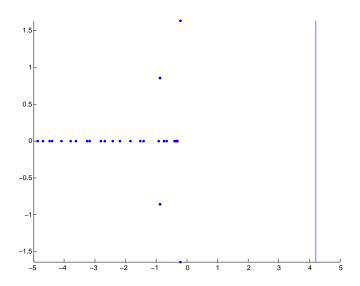
$$\frac{d}{dt}||u||^2=2u^TT_{22}u.$$

Define spectral abscissa

$$\omega(T_{22}) = \max_{v} \{ v^{T} T_{22} v \} = \lambda_{\max}(H(T_{22}))$$

Finding  $\omega(T_{22})$  is a symmetric exterior eigenvalue problem!  $\implies$  estimate with Lanczos.

## Bound applied to a 2D Brusselator



### Conclusions

- Continuing eigendecompositions is useful for resonator design and for bifurcation analysis
- ▶ Basic algorithm: predictor-corrector + Krylov subspaces
- Tests to make sure computed subspace is good enough
- Ongoing software work (CL-MATCONT extension)