Koopman modes, Resolvent modes and a lid-driven cavity

Recurrent Flows, The Clockwork Behind Turbulence Kavli Institute of Theoretical Physics University of California, Santa Barbara

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cavity / pipe DMD: P Gomez, H Blackburn, M Rudman (Monash), B McKeon (Caltech)

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Why do DMD modes look like resolvent modes?



Gomez et al PoF 2014

"Paco's pipe projection puzzle"

n = 2, $\omega/2\pi = 0.1826$, 0.5497, isosurfaces $1/3u_{max}$.

Koopman modes are a general way of analysing nonlinear system dynamics, closely related to DMD*.

Why do resolvent modes look like DMD/Koopman modes?

Can we determine Koopman modes with limited data?

*Rowley et al JFM 2009; Schmid JFM 2010

The Koopman operator

Suppose the dynamical system with state **u**,

 $\partial_t \mathbf{u}(x,t) = \mathbf{f}(\mathbf{u}(x,t)).$

Define a state transformation

 $S^{\tau}(\mathbf{u}(x,t)) = \mathbf{u}(x,t+\tau).$

A Koopman operator \mathbf{U}^{τ} is defined for each S^{τ} ,

 $\mathbf{U}^{\tau}h(\mathbf{u}) = h(S^{\tau}\mathbf{u}).$

The (family of) Koopman operators $\{U^{\tau}\}$ describes the dynamics of observable $h(\mathbf{u})$.

Spectrum of the Koopman operator

Consider the eigenvalue problem for \mathbf{U}^{τ}

 $\mathbf{U}^{\tau}\varphi_n(\mathbf{u}) = \mu_n \varphi_n(\mathbf{u}).$

Expanding some vector valued observable $\mathbf{g}(\mathbf{u})$ accordingly,

$$\mathbf{g}(\mathbf{u}) = \sum_{n} \varphi_n(\mathbf{u}) \mu_n \mathbf{g}_n$$

Mezic Nonlin. Dyn. 2005, A. Rev. Fl. Mech. 2013; Rowley et al, JFM 2009

Koopman modes

For flow on a periodic orbit, decomposition in these eigenfunctions and Koopman modes reduces to a Fourier decomposition,

$$\mathbf{g}(x,t) = \mathbf{g}^*(x) + \sum_{n \neq 0} \mathbf{g}_n(x) e^{in\omega t}.$$

The Koopman mode $\mathbf{g}_n(x)$ is the projection of the field \mathbf{g} on the subspace spanned by an eigenfunction of \mathbf{U}^{τ} .

Note: temporal average emerges as $\mathbf{g}^*(x)$.

Koopman mode interpretation

Points to note:

- ► Selects frequencies present in system behaviour
- Chaotic system may have point and continuous spectrum
- ► Spatial structure is wrapped up in Koopman mode
- ► Periodic orbit gives Fourier decomposition
- ► ...but any transformation (e.g. reflection, rotation) also possible

Generalising Koopman operator

We can define the Koopman modes where the Koopman operator does another transformation (spatial shift, reflection, etc)

Define a state transformation

 $S^{\chi}(\mathbf{u}(x,t)) = \mathbf{u}(x+\chi,t).$

When periodic in x, \mathbf{U}^{χ} gives

$$\mathbf{g}(x,t) = \bar{\mathbf{g}}(t) + \sum_{l \neq 0} \mathbf{g}_l(t) e^{il\alpha x}.$$

The spatial Koopman modes are $\mathbf{g}_l(t)$ Note: spatial average appears as $\bar{\mathbf{g}}(t)$.

Sharma, Mezic, McKeon, PRF, 2016

Combining Koopman operators

Combining the Koopman operators,

 $\mathbf{U}^{\chi}\mathbf{U}^{\tau}\mathbf{g}(\mathbf{u}) = \mu\mathbf{g}(\mathbf{u}).$

gives

$$\mathbf{g}(x,t) = \sum_{n,m\in\mathbb{Z}} \mathbf{g}_{n,m} e^{i(n\omega t + m\alpha x)},$$

which is an expansion in travelling waves.

For turbulent attractor, same applies but continuous instead of point spectrum.

An expansion around the turbulent attractor I

Let $\mathbf{u}(t)$ be the state, and the Navier-Stokes equations be written

$$\dot{\mathbf{u}}(t) = \mathbf{f}(\mathbf{u}(t)). \tag{1}$$

In the "long run", decompose the state as

$$\mathbf{u}(t) = \frac{1}{2\pi i} \int_{-\infty}^{\infty} e^{i\omega t} \hat{\mathbf{u}}(\omega) d\omega.$$

Notice that the equation corresponding to $\omega = 0$ is the mean equation

$$0 = \int_{-\infty}^{\infty} \mathbf{f}(\mathbf{u}) dt$$

with $\hat{\mathbf{u}}(0)$ the mean.

An expansion around the turbulent attractor II

The expansion of (1) about this mean (and subtracting this mean equation) is

$$\dot{\tilde{\mathbf{u}}}(t) = \frac{\partial \mathbf{f}}{\partial \mathbf{u}} \Big|_{\bar{\mathbf{u}}} \tilde{\mathbf{u}}(t) + \mathcal{O}(2)$$
$$= \mathbf{L}\tilde{\mathbf{u}}(t) + \tilde{\mathbf{g}}(t)$$

where $\tilde{\mathbf{g}}$ represents the second-order terms in the expansion of \mathbf{f} .

At a particular $\omega \neq 0$ we then have

$$\hat{\mathbf{u}}(\omega) = (\omega I - \mathbf{L})^{-1} \hat{\mathbf{g}}(\omega).$$

The second-order terms, rather than being truncated, act to excite the state.

A low-rank basis

The resolvent $\mathbf{H}(\omega) = (\omega I - \mathbf{L})^{-1}$ is well-approximated by a projection $\Pi(\omega)$,

 $\hat{\mathbf{u}}(\omega) = \mathbf{H}(\omega)\hat{\mathbf{g}}(\omega) \simeq \Pi(\omega)\hat{\mathbf{g}}(\omega).$

The SVD gives the optimal *N*-dimensional basis on which the velocity field evolves, in sense that $\|\mathbf{H}(\omega) - \Pi_N(\omega)\|_F$ is minimised. ^[1]

The flow "lives" mostly in Π .

The 'error' is $\hat{\mathbf{u}}^{\perp} = \Pi_N^{\perp} \hat{\mathbf{g}}$.

[1] McKeon & Sharma, JFM, 2010.

A basis for optimal projection; resolvent modes

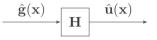
$$\mathbf{H}\hat{\mathbf{g}} = \sum_{m=1}^{\infty} \psi_m(\mathbf{x}) \sigma_m \left\langle \phi_m^*(\mathbf{x}), \hat{\mathbf{g}}(\mathbf{x}) \right\rangle$$

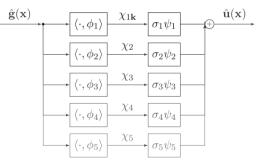
$$\langle \psi_m, \psi_{m'} \rangle = \delta_{m,m'}$$

$$\langle \phi_m, \phi_{m'} \rangle = \delta_{m,m'}$$

$$\sigma_1 \ge \sigma_2 \ge \dots$$

Each σ_m is a (real) gain, σ_1 is the maximum gain. Velocity field response is $\psi_m(\mathbf{x})$.





Relation between Koopman and resolvent modes

Comparing starting point of RM derivation, we see that the resolvent relates the Koopman modes of ${\bf u}$ and ${\bf f}$

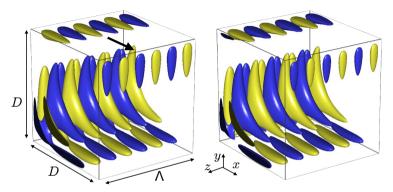
 $\mathbf{u}_m(y) = \mathcal{H}_m \mathbf{f}_m(y)$

so the resolvent modes are (linear-)optimal basis to guess the Koopman modes.

Interpretation

- ► Resolvent modes a good 'first guess' of KM; for ordered expansion of KMs
- Koopman modes \equiv RM modes \equiv Fourier in homogeneous directions
- ► Where there are symmetries, they should be used *before* doing the DMD problem
- Expansion in RMs easily gives sensitivity of KMs to control / BC changes
- Meaning of spectrum of linear operator $\mathcal{H}_m(\mathbf{u}_0)$ is clear

Application: 2D resolvent modes in a lid-driven cavity



- ► few, discrete frequencies
- ► complex geometry

Lid-driven cavity

- $\blacktriangleright \ Re = 1200, \, \Lambda = 0.945 D$
- ► Nonlinear, low-dimensional behaviour
- Three dominant wavenumbers: $\beta = 0, 3, 6$
- Three dominant frequencies: $\omega = 0, 0.76, 1.52$
- ► spectral-*hp* 2D × Fourier (semtex)

Fourier transform velocity (translation invariant in time, z; neglect transients)

$$\mathbf{u} = \sum_{\beta,\omega} \mathbf{u}_{\beta,\omega}(x,y) \mathrm{e}^{i(\beta z - \omega t)}$$

Assume time-space mean \mathbf{u}_0 to close.

Same for nonlinear term,

$$-\mathbf{u} \cdot \nabla \mathbf{u} = \sum_{\beta,\omega} \mathbf{f}_{\beta,\omega}(x,y) \, \mathrm{e}^{i(\beta z - \omega t)}$$

Fluctuations:

$$\mathbf{u}_{\beta,\omega} = (i\omega - \mathcal{L}_{\beta})^{-1} \mathbf{f}_{\beta,\omega}.$$

Mean:

$$0 = \mathbf{f}_0 - \mathbf{u}_0 \cdot \nabla \mathbf{u}_0 + \frac{1}{Re} \nabla^2 \mathbf{u}_0.$$

Take SVD of transfer function,

$$(i\omega - \mathcal{L}_{\beta})^{-1} = \sum_{m} \psi_{\beta,\omega,m} \sigma_{\beta,\omega,m} \phi^*_{\beta,\omega,m}$$

Gives gain-optimal basis to represent \mathbf{u} and \mathbf{f} , scalar coefficient c,

$$\mathbf{u}_{\beta,\omega}(x,y) = \sum_{m} \psi_{\beta,\omega,m}(x,y) \ c_{\beta,\omega,m}$$

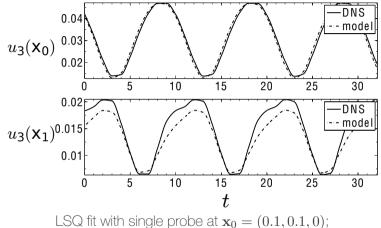
$$\mathbf{f}_{\beta,\omega}(x,y) = \sum_{m} \phi_{\beta,\omega,m}(x,y) \ c_{\beta,\omega,m} \ / \ \sigma_{\beta,\omega,m}$$

Estimating mode coefficients from probe signal

Focus on $\beta = 3$, fit m = 1 coefficients at three frequencies.

 $C_{\beta} = \Psi_{\beta}^{+}(\mathbf{x}_{p})U_{\beta}(\mathbf{x}_{p})$

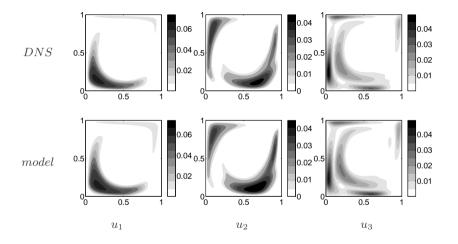
Reconstructed signals



reconstruction at \mathbf{x}_0 and $\mathbf{x}_1 = (0.82, 0.95, 0)$

Reconstructed field (RMS fluctuations)

/



About 20% model error

isosurfaces at 30% max $w_{\beta=3}$.

Conclusions (cavity)

Limitations:

▶ Needs limited data to fix amplitudes, phases, dominant ω , β , mean

Benefits:

- Meaning of mean flow in linear operators is now clear
- Approximates whole flow from probed points + mean
- Modes are orthogonal (unlike eigenmodes)

Conclusions

- ► DMD modes look like RM because RM look like Koopman modes
- ► Resolvent modes offer (linear-) optimal basis to expand Koopman modes of velocity
- ► For homogeneous directions, it's just Fourier modes
- ► Global coherence allows fixing of coefficients from point data
- ► Can use to predict response of nonlinear system to control