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$$A_i - \sigma_i I = Q_i R_i$$

$$A_{i+1} = R_i Q_i + \sigma_{i+1} I$$

Projected Eigensolvers:

Galerkin: $Au - \lambda u \perp K$

$$(Au - \lambda u, v) = 0 \quad \forall v \in K$$

Suppose that $\underbrace{\{v_1, \dots, v_m\}}_V$ is an orthonormal basis for K .

Then,

$$u = y_1 v_1 + y_2 v_2 + \dots + y_m v_m = Vy$$

$$(AVy - \lambda Vy, v_j) = 0 \quad j=1, \dots, m$$

$$V^T AVy = \lambda y$$

Recall

$$Q^T A Q = \begin{cases} T \\ H \end{cases} \begin{array}{l} \text{- Lanczos} \\ \text{- Arnoldi} \end{array}$$

$$Q_k = \begin{bmatrix} \vdots \\ \vdots \\ \vdots \end{bmatrix}_n^k$$

$$\underbrace{Q_k^T A Q_k}_{\substack{T_k \\ H_k}} \in \mathbb{R}^{k \times k}$$

Ritz values:

eigenvalues of T_k or H_k

\hookrightarrow approx eig of A

$$\left. \begin{array}{l} AV_m = V_m H_m + h_{m+1,m} V_{m+1} e_m^T \\ V_m^T AV_m = H_m \end{array} \right\} \text{Arnoldi}$$

$$H_m y_i^{(m)} = \lambda_i^{(m)} y_i^{(m)}$$

$\lambda_i^{(m)}$ Ritz values } approximations
 $u_i^{(m)} = V_m y_i^{(m)}$ Ritz vectors } for eigenpairs.
 ↑ Rayleigh-Ritz.

recall $(Q^T A Q)(Q^T x) = \lambda (Q^T x)$

$$A V_m y_i^{(m)} = \underbrace{V_m H_m y_i^{(m)}}_{\lambda_i^{(m)} y_i^{(m)}} + h_{m+1,m} V_{m+1} e_m^T y_i^{(m)}$$

$$A V_m y_i^{(m)} - \lambda_i^{(m)} V_m y_i^{(m)} = h_{m+1,m} e_m^T y_i^{(m)} V_{m+1}$$

$$\|(A - \lambda_i^{(m)} I) u_i^{(m)}\|_2 = h_{m+1,m} |e_m^T y_i^{(m)}|$$

cheap, computable way to estimate residual

Lanczos:

$$T = V^T A V = \begin{bmatrix} V_m & V_u \end{bmatrix}^T A \begin{bmatrix} V_m & V_u \end{bmatrix} = \begin{pmatrix} V_m^T A V_m & V_m^T A V_u \\ V_u^T A V_m & V_u^T A V_u \end{pmatrix}$$

where $V = \begin{bmatrix} V_m & V_u \end{bmatrix}$

$$= \begin{pmatrix} T_m & T_{mu} \\ T_{mu}^T & T_{uu} \end{pmatrix}$$

Thm: the minimum of

$$\|A V_m - V_m R\|_2$$

over all $m \times m$ symmetric matrices R is attained by $R = T_m$.

$$\|A V_m - V_m R\|_2^2 = \lambda_{\max}((A V_m - V_m R)^T (A V_m - V_m R))$$

write $R = T + Z$

show that $Z = 0$ minimizes this expression.

recall $\|B\|_2 = \max_{x \neq 0} \frac{\|Bx\|_2}{\|x\|_2} = \sigma(B)$
 $= \max_i |\lambda_i(B^T B)|$

drop subscripts:

$$\begin{aligned}
 & (AV - V(T+Z))^T (AV - V(T+Z)) \quad \begin{matrix} 0 & \text{since } V^T A V = T & 0 \end{matrix} \\
 & = (AV - VT)^T (AV - VT) - \underbrace{(AV - VT)^T V Z}_{0} - \underbrace{(VZ)^T (AV - VT)}_{0} + \underbrace{(VZ)^T (VZ)}_{Z^T Z = \text{SPSD}}
 \end{aligned}$$

Since $Z^T Z$ SPBD

it can only increase the eigenvalues, therefore, take $Z=0$

$$\text{Also } (V^T Z)^T (AV - VT) = Z^T (V^T AV - T) \\ = 0$$

$\begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} T$ is a minimizer

1st of accomplishments

- algorithm: Arnoldi + Lanczos
- approx: Ritz values + vectors
- optimality result: best in least squares sense (generalizes R.Q.)

$$x_q = P_q(A) x_0 = \sum_{i=1}^n p_q(\lambda_i) \gamma_i u_i$$

↑ polynomial

$$\text{recall } A^k u_i = \lambda_i^k u_i$$

$$\{(\lambda_i, u_i)\}_{i=1}^n \text{ eigenpairs}$$

$$x_0 = \sum_{i=1}^n \gamma_i u_i$$

$$x_q = \sum_{i=1}^n p_q(\lambda_i) \gamma_i u_i = p_q(\lambda_1) \gamma_1 u_1 + \sum_{i=2}^n p_q(\lambda_i) \gamma_i u_i$$

Suppose we want the first eigenpair:

$$p_q(\lambda_1) \text{ large}$$

$$p_q(\lambda_i) \text{ small } i > 1$$

We don't know the eigenvalues, or the polynomial.

We do have Ritz values.

$$\text{So: } p(t) = (t - \theta_2)(t - \theta_3) \dots (t - \theta_k)$$

ϵ small near $\theta_2, \theta_3, \dots, \theta_k$

ϵ zero at $\theta_2, \theta_3, \dots, \theta_k$

\hookrightarrow hopefully small near $\lambda_2, \lambda_3, \dots, \lambda_k$

restarting:

- can be done simply by taking the desired Ritz vector so far, reset $m=1$, take it and use it as an initial guess for a fresh Arnoldi.

ARPACK