## CS556 Iterative Methods Fall 2020 Eigenvalues from PCG

Consider the following PCG algorithm with A and preconditioner M SPD.

for 
$$k = 1:n$$
 (1)  
Solve  $Mr = r = 0$   $r^T r = \theta = \theta k$  if  $k = 1, \theta = 0$  (2)

Solve 
$$M \underline{z}_k = \underline{r}_{k-1}, \ \rho_k = \underline{z}_k^I \underline{r}_{k-1}, \ \beta_k = \frac{\rho_k}{\rho_{k-1}}, \ \text{if } k=1, \ \beta_k = 0$$
 (2)

$$\underline{p}_k = \underline{z}_k + \beta_k \underline{p}_{k-1} \tag{3}$$

$$\underline{w}_k = A\underline{p}_k, \ \gamma_k = \underline{p}_k^T \underline{w}_k, \ \alpha_k = \frac{\rho_k}{\gamma_k}$$
(4)

$$\underline{x}_k = \underline{x}_{k-1} + \alpha_k \underline{p}_k \tag{5}$$

$$\underline{r}_k = \underline{r}_{k-1} - \alpha_k \underline{w}_k \tag{6}$$

end 
$$(7)$$

Let

$$R = [\underline{r}_0 \ \underline{r}_1 \ \dots \ \underline{r}_{k-1}], \tag{8}$$

$$P = [\underline{p}_1 \ \underline{p}_2 \ \dots \ \underline{p}_k], \tag{9}$$

$$Z = [\underline{z}_1 \ \underline{z}_2 \ \dots \ \underline{z}_k], \tag{10}$$

and

$$B = \begin{bmatrix} 1 & -\beta_2 & & \\ & 1 & -\beta_3 & \\ & & 1 & -\beta_4 \\ & & & \ddots \end{bmatrix}.$$
(11)

Note that  $P^T A P = \text{diag}(\gamma_i)$ . We also have that Z = PB, so

$$Z^T A Z = \tilde{T}, \tag{12}$$

where  $\tilde{T}$  is a  $k \times k$  tridiagonal matrix. Moreover,  $Z^T M Z = Z^T R = \text{diag}(\rho_i) =: \Delta^2$ . From these, we can find approximate eigenvectors and eigenvalues for the generalized eigenvalue problem,  $A\underline{s}_j = \lambda_j M\underline{s}_j$ . Let's consider  $\lambda_n$ , the eigenvalue that maximizes the Rayleigh quotient,

$$\lambda_n = \max_{\underline{s} \in \mathbb{R}^n} \frac{\underline{s}^T A \underline{s}}{\underline{s}^T M \underline{s}}.$$
(13)

The approximated value is

$$\lambda_n \approx \max_{\underline{z} \in Z} \frac{\underline{z}^T A \underline{z}}{\underline{z}^T M \underline{z}} = \max_{\underline{y} \in \mathbb{R}^k} \frac{\underline{y}^T Z^T A Z \underline{y}}{\underline{y}^T Z^T M Z \underline{y}} = \max_{\underline{y} \in \mathbb{R}^k} \frac{\underline{y}^T \tilde{T} \underline{y}}{\underline{y}^T \Delta^2 \underline{y}} = \mu_k, \tag{14}$$

where  $\mu_k$  is the maximum eigenvalue for the  $k \times k$  generalized eigenvalue problem,

$$\tilde{T}\underline{y}_{j} = \mu_{j}\Delta^{2}\underline{y}_{j} \tag{15}$$

Here,  $\Delta^2$  is a diagonal matrix. Define  $\underline{u} = \Delta \underline{y} \longrightarrow \underline{y} = \Delta^{-1} \underline{u}$ . The right-most Rayleigh quotient in the preceding equation becomes

$$\mu_k = = \max_{\underline{u} \in \mathbb{R}^k} \frac{\underline{u}^T T \underline{u}}{\underline{u}^T \underline{u}},$$
(16)

where  $T := \Delta^{-1} \tilde{T} \Delta^{-1}$ .