## Sensitivity of Simple Eigenvalues

Let  $A \in \mathbb{C}^{n \times n}$ . We would like to know how small perturbations in A change its eigenvalues. Of course,  $\partial \lambda_k / \partial a_{ij}$  measures just this sensitivity, but it isn't practical to compute these  $n^3$  quantities. Suppose  $Ax = \lambda x$ ,  $y^*A = \lambda y^*$ , and  $||x||_2 = 1 = ||y||_2$ (so x and  $y^*$  are respectively right and left eigenvectors associated with a simple (not multiple) eigenvalue  $\lambda$  of A). Remember,  $y^* = \overline{y}^T$  is a conjugate transpose.

Consider the perturbed eigenpair x(t),  $\lambda(t)$  of the matrix A + tE:

$$(A + tE)x(t) = \lambda(t)x(t),$$

where x(0) = x,  $\lambda(0) = \lambda$ , and  $||x(t)||_2 = 1$ . Differentiating wrt t gives:

$$(A + tE)\dot{x}(t) + Ex(t) = \lambda \dot{x}(t) + \dot{\lambda}x(t).$$

Premultiplying by  $y^*$  gives  $\dot{\lambda}(0) = y^* E x / (y^* x)$ , and for  $\lambda \neq 0$ , Taylor's theorem says

$$\begin{array}{rcl} \frac{\lambda(t)-\lambda}{\lambda}| &=& |\frac{t\dot{\lambda}}{\lambda}+\mathcal{O}(t^2)|\\ &\approx& |\frac{ty^*Ex}{\lambda y^*x}|\\ &\leq& \frac{\|A\|}{|\lambda y^*x|}\frac{\|tE\|}{\|A\|}. \end{array}$$

Thus we say the relative (and absolute) condition numbers for  $\lambda$  are

$$\kappa(\lambda) = \frac{\|A\|}{|\lambda|} \frac{1}{|y^* x|} \qquad (\text{ and } \qquad \nu(\lambda) = \frac{1}{|y^* x|}).$$

So, if  $y^*x$  is small,  $\lambda$  is illconditioned. If  $\lambda$  is simple,  $y^*x$  cannot be zero, but for some matrices it can be very small. Let's take a diagonalizable matrix under consideration. Suppose  $X^{-1}AX = \Lambda = \text{diag}(\lambda_1, \ldots, \lambda_n)$ . The columns of X are (normalized) right eigenvectors of A, while the rows of  $X^{-1} = DY$  are (un-normalized) left eigenvectors. Let  $D = \text{diag}(||e_i^T X^{-1}||, i = 1, \ldots, n)$ , so that the rows of Y have unit length. Then from above,

$$\nu(\lambda_i) = |e_i^T Y X e_i|^{-1} = d_{ii}.$$

Therefore, if  $||X|| ||X^{-1}||$  is large, then at least one eigenvalue of A is illconditioned. Conversely, if  $||X|| ||X^{-1}||$  is not large, then all eigenvalues of A are well conditioned. Some eigenvalues of A may be sensitive, while others may not, but to describe the absolute sensitivity of the eigenvalues of A with a single number, a good choice is

$$\nu_{eig}(A) = \kappa_{inv}(X) = \|X\| \|X^{-1}\|.$$

A matrix A is called *normal* if  $AA^* = A^*A$ . Hermitian matrices are normal, as are unitary and diagonal matrices. The eigenvectors of normal matrices can always be arranged to be orthonormal, so if A is normal, then  $\nu_{eig}(A) = 1$ . The farther from normal a matrix is, the more sensitive (illconditioned) are its eigenvalues.