

# 1. Eigenvalue algorithm

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Given a symmetric matrix  $A \in \mathbb{R}^{n \times n}$  with  $n$  different eigenvalues, write a program to compute its eigenvalues and eigenvectors in Armadillo. Equivalently, approximately find its eigendecomposition:

$$A = \sum_{i=1}^n \lambda_i u_i u_i^T, \quad (\lambda_i \neq \lambda_j, \forall i \neq j)$$

You can use **power iteration algorithm** as follows:

Starts with a vector  $b_0$ . The method is described by the recurrence relation

$$b_{k+1} = \frac{Ab_k}{\|Ab_k\|}$$
$$\mu_k = b_k^T Ab_k$$

Under some conditions,  $\mu_k, b_k$  converges to the dominant eigenvalue and corresponding eigenvector.

Check your algorithm by generating a  $3 \times 3$  random matrix  $A$  with standard normal entries and then set  $A = A + A^T$ . Compare your result with `eig_sym()` in Armadillo.

# 2. Generating gaussian distribution

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(a)

Write a program to generate standard normal sample using uniform random sample generator(for example, `randu()` in Armadillo or `uniform_dist` in `trng`). This can be done by following **Box-Muller transform**:

Given  $U_1$  and  $U_2$  independently from `Uniform(0, 1)`.

Consider polar coordinates  $(R, \Theta)$  for independent standard normal  $(Z_0, Z_1)$ . we know  $R^2 \sim \chi_2^2$  and  $\Theta \sim \text{Uniform}(0, 2\pi)$ , thus generate  $R$  and  $\Theta$  by:

$$R = \sqrt{-2 \ln U_1}$$
$$\Theta = 2\pi U_2$$

Then independent standard normal  $(Z_0, Z_1)$  can be obtained by

$$Z_0 = R \cos(\Theta) = \sqrt{-2 \ln U_1} \cos(2\pi U_2)$$
$$Z_1 = R \sin(\Theta) = \sqrt{-2 \ln U_1} \sin(2\pi U_2)$$

(b)

Why your generated  $Z_0$  and  $Z_1$  are actually truncated? (hint: for a computer, there is a smallest non-zero number)

