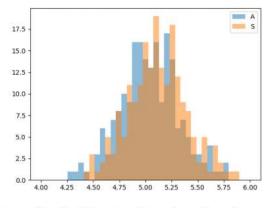
Histogram of max eigenvalue values for 10x10 random matrices A and  $S = \frac{1}{2}(A + A^T)$ 



It seems like the histogram of max eigenvalue values represents some distribution where 5.00 is the mean, and the values overall fall in a range not exceeding 6.00 or falling below 4. The code is below.

```
import numpy as np
from numpy import linalg as LA
import matplotlib.pyplot as plt
def generate_matrix ():
         mat = np.random.rand(10,10)
         return mat
def get_max_eigenvalue(mat):
         w, v = LA. eig(mat)
         return np.amax(w)
def run_problem ():
         \begin{array}{l} A\_maxes = []\\ S\_maxes = [] \end{array}
         for i in range (200):
                   A = generate_matrix()
                   S = 0.5 * (A + np.transpose(A))
                   A_maxes.append(get_max_eigenvalue(A))
                   S_maxes.append(get_max_eigenvalue(S))
         return [A_maxes, S_maxes]
A_maxes, S_maxes = run_problem()
np. histogram (A_maxes)
bins = np.linspace(4, 6, 40)
plt.hist(A_maxes, bins, alpha=0.5, label='A')
plt.hist(S_maxes, bins, alpha=0.5, label='S')
plt.legend(loc='upper_right')
plt.show()
```

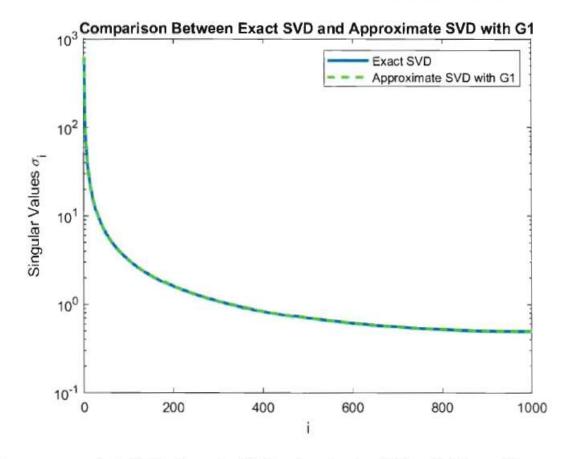
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2.

```
A = triu(ones(1000));
G1 = normrnd(0,1,1000,10); % 1000 by 10 Gaussian random matrix G1
G2 = normrnd(0,1,1000,100); % 1000 by 100 Gaussian random matrix G2
% Exact or Actual SVD
[u,s,v] = svd(A);
% Randomized or Approximate SVD with G1
Y1 = A*G1;
[Q1,R1] = qr(Y1);
[U1,D1,V1] = svd(Q1'*A); W1 = Q1*U1; % A = (Q1*U1)*D1*V1' = W1*D1*V1
% Randomized or Approximate SVD with G2
Y2 = A^*G2;
[Q2, R2] = qr(Y2);
[U2,D2,V2] = svd(Q2'*A); W2 = Q2*U2; % A = (Q2*U2)*D2*V2' = W2*D2*V2'
s11 = diag(s); display(s11(1:10)) % The 10 Largest Singular Values From Actual SVD
           636.93814767091
          212.312890336131
          127.387943537033
          90.9916125292626
          70.7714867858701
          57.9041816178516
          48,9960875295752
          42.4635200898666
          37.4680581293865
           33.524299918613
D11 = diag(D1); display(D11(1:10)) % The 10 Largest Singular Values From Approximate SVD with G1
          636.938147670908
           212.312890336131
          127.387943537033
          90.9916125292626
          70.7714867858701
          57.9041816178517
          48.9960875295753
           42.4635200898666
          37.4680581293866
           33.524299918613
D22 = diag(D2); display(D22(1:10)) % The 10 Largest Singular Values From Approximate SVD with G2
           636.938147670908
           212.31289033613
           127.387943537033
           90.9916125292626
           70.7714867858701
           57.9041816178516
          48.9960875295752
           42.4635200898666
           37,4680581293866
           33.524299918613
```

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```
% Comparison Between Actual SVD and Randomized SVD with G1
% Plot of the Singular Values of the Actual SVD and Randomized SVD with G1
figure();
semilogy(s11,'linewidth',2)
hold on
semilogy(D11,'--g','linewidth',1.5)
grid on
title('Comparison Between Exact SVD and Approximate SVD with G1')
legend('Exact SVD','Approximate SVD with G1','location','Northeast')
xlabel('i')
ylabel('Singular Values \sigma_i')
```



As we can see, both SVD's, the actual SVD and randomized SVD with G1, result in approximately same Singular Values.