

$\Rightarrow H$ 对称正定

3.31. $Ax \leq b$ 的最小二乘解 $x = A^+b = V\Sigma^+U^Tb$.

$$\Rightarrow x = V\Sigma^+(U_1^Tb, U_2^Tb, \dots, U_m^Tb)$$

$$= (V_1, V_2, \dots, V_m) \begin{pmatrix} \frac{U_1^Tb}{\sigma_1}, & \dots, & \frac{U_m^Tb}{\sigma_m} \end{pmatrix}, \sigma_i \geq 0.$$

$$= \sum_{\sigma_i \geq 0} \frac{U_i^Tb}{\sigma_i} V_i$$

(由于 $\sigma_i \neq 0$ 时, 对应广义逆为 0).

3.32. (a) $AA^+A = A \Leftrightarrow V\Sigma\Sigma^+\Sigma V^T = V\Sigma V^T$.

由于 $\Sigma\Sigma^+\Sigma = \Sigma^*$. \Rightarrow 结论成立.

(b). $\Sigma^+\Sigma\Sigma^+ = \Sigma^+$. $\Rightarrow A^+AA^+ = A^+$.

(c) 由于 $\Sigma\Sigma^+$ 为对角阵. $\Rightarrow (\Sigma\Sigma^+)^T = (\Sigma\Sigma^+)$

\Rightarrow 结论成立

(d) 同上, 结论成立.

Hw4

4.13. 设 $Ax_0 = \lambda_{\max}x_0$, $\|x_0\| = 1$

$\Rightarrow \|A\| = \sup_{\|x\|=1} \|Ax\| \geq \lambda_{\max}$. 即 $\rho(A) \leq \|A\|$

4.19 $Ax = \lambda x$, $\Rightarrow \lambda^2 x = A^2 x = Ax = \lambda x$

$\Rightarrow \lambda = 1$ 或 0.

4.20. (a) $Ax = \lambda x$, $Ay = \mu y$. $\Rightarrow y^H A x = \lambda y^H x$.

$$x^H A y = \lambda x^H y$$

$$(b) \Rightarrow y^H x = 0.$$

$$y^H A x = \lambda y^H x$$

另一方面, $y^H A x = y^H (\lambda x) = \lambda y^H x$.

(c) 考查 A 的 Jordan 标准型.

\exists 可逆矩阵 P , st $P^{-1}AP = [\lambda_j]$

$\Rightarrow AP = P[\lambda_j] \Rightarrow P$ 的第一列为 λ 的右特征值

$P^{-1}A - [\lambda_j]P^{-1}, \Rightarrow P^{-1}$ 的第一行为 λ 的左特征值

由于 $P^{-1}P = I \Rightarrow y^H x \neq 0$

4.24

$$(a) A = U \Sigma V^T = U \begin{pmatrix} \sqrt{\sigma_0} & & \\ & \ddots & \\ & & \sqrt{\sigma_n} \end{pmatrix} \begin{pmatrix} \sqrt{\sigma_0} & & \\ & \ddots & \\ & & \sqrt{\sigma_n} \end{pmatrix} V^T$$

$$= \sqrt{\sigma} U (\sqrt{\sigma} V)^T$$

令 $U = \sqrt{\sigma} U_1, V = \sqrt{\sigma} V_1$, 即可.

$$(b) AU = UV^T U = (V^T U) \cdot U \Rightarrow V^T U = U^T V \text{ 的特征值}$$

(c) 0.

$$(d) \forall X, Ax = UV^T X = U(V^T X), \Rightarrow \text{一步}$$

4.25. 由于 $r(A) = 1 \Rightarrow \exists$ 正交阵 V , st $U^T \begin{pmatrix} \sigma_0 & & \\ & \ddots & \\ & & 0 \end{pmatrix} V = A$.

$$\Rightarrow \det(I + UV^T) = 1 + \sigma = 1 + U^T V.$$

Exercise

上海学生统一课业簿册

4.27. (a) 如 $(I - A)x = 0$, 则 $Ax = x$.

$\Rightarrow \text{PCA} \geq I$, (A 有 1 特征值)

矛盾.

$$(b) \left(\sum_{k=0}^{\infty} A^k \right) (I - A) = \sum_{k=0}^{\infty} (A^k - A^{k+1}) = I$$

$$(I - A) \left(\sum_{k=0}^{\infty} A^k \right) = \sum_{k=0}^{\infty} (A^k - A^{k+1}) = I.$$

$$\Rightarrow (I - A)^{-1} = \sum_{k=0}^{\infty} A^k.$$

4.31 (a) 全 $Ax = \lambda x$, $\Rightarrow (Ax)^T (Ax) = x^T x$.

又由于 $(Ax)^T Ax = \lambda^2 x^T x \Rightarrow \lambda^2 = 1 \Rightarrow |\lambda| = 1$.

(b) 由于 $A^T A = I$, \Rightarrow 其奇异值为 ± 1 .

4.32. (a) $H = H^T = H^{-1}$, $\Rightarrow |\lambda| = 1$, $\lambda = 1$, 1 重数为 1, 1 的重数为 $n-1$.

$$(b) a = \begin{pmatrix} c & s \\ -s & c \end{pmatrix}.$$

$$\begin{vmatrix} \lambda - c & s \\ -s & \lambda - c \end{vmatrix} = (\lambda - c)^2 + s^2 = 0, \Rightarrow \lambda = c \pm is.$$

$$5.3. (a) x_{n+1} = \frac{x_{n-1} f(x_n) - x_n f(x_{n-1})}{f(x_n) - f(x_{n-1})} = \frac{f(x_n) x_{n+1} - x_n f(x_n) + x_n f(x_n)}{f(x_n) - f(x_{n-1})} \\ = x_n - \frac{x_n - x_{n-1}}{f(x_n) - f(x_{n-1})}.$$

(b) 当 $f(x_n) \rightarrow 0$ 时, $f(x_n) - f(x_{n-1}) \rightarrow 0$, 牛顿法误差可能
较大. 而在 (a) 中, 有 $f(x_n) \approx f(x_{n-1})$ 时 $x_n \approx x_{n-1}$,
使误差变小.

T1

Computer problem (in C or C++): Write the functions to achieve (1) Arnoldi iteration and Lanczos iteration; (2) QR iteration for Hessenburg matrix to find the eigenvalue decomposition. Test these algorithms for a few matrix.

Arnoldi iteration

In []:

```
import numpy as np
import matplotlib.pyplot as plt
from scipy.linalg import circulant, solve

def arnoldi_iteration(A, b, n: int):
    """Computes a basis of the (n + 1)-Krylov subspace of A: the space
    spanned by {b, Ab, ..., A^n b}.

    Arguments
    A: m × m array
    b: initial vector (length m)
    n: dimension of Krylov subspace, must be >= 1

    Returns
    Q: m × (n + 1) array, the columns are an orthonormal basis of the
        Krylov subspace.
    h: (n + 1) × n array, A on basis Q. It is upper Hessenberg.
    """
    eps = 1e-12
    h = np.zeros((n+1, n), dtype=np.complex)
    Q = np.zeros((A.shape[0], n+1), dtype=np.complex)
    # Normalize the input vector
    Q[:, 0] = b / np.linalg.norm(b, 2) # Use it as the first Krylov vector
    for k in range(1, n+1):
        v = np.dot(A, Q[:, k-1]) # Generate a new candidate vector
        for j in range(k): # Subtract the projections on previous vectors
            h[j, k-1] = np.dot(Q[:, j].conj(), v)
            v = v - h[j, k-1] * Q[:, j]
        h[k, k-1] = np.linalg.norm(v, 2)
        if h[k, k-1] > eps: # Add the produced vector to the list, unless
            Q[:, k] = v / h[k, k-1]
        else: # If that happens, stop iterating.
            return Q, h
    return Q, h
```

Lanczos iteration

In []:

```
def lanczos_iteration(A, b, n: int):
    eps = 1e-12
    beta = 0
    Q = np.zeros((A.shape[0], n+2), dtype=np.complex)
    h = np.zeros((n+1, n), dtype=np.complex)
    # Normalize the input vector
    Q[:, 1] = b / np.linalg.norm(b, 2) # Use it as the first Krylov vector
    for k in range(2, n+2):
        v = np.dot(A, Q[:, k-1]) # Generate a new candidate vector
        alpha = Q[:, k-1].conj().T @ v
        v = v - beta * Q[:, k-2] - alpha * Q[:, k-1]
        beta = np.linalg.norm(v, 2)
        h[k-2, k-2] = alpha
        h[k-1, k-2] = beta
        if k < n+1:
            h[k-2, k-3] = beta
        if beta > eps: # Add the produced vector to the list, unless
            Q[:, k] = v / beta
        else: # If that happens, stop iterating.
            return Q[:, 1:], h
    return Q[:, 1:], h
```

In []:

```
N = 2**4
I = np.eye(N)
k = np.fft.fftfreq(N, 1.0 / N) + 0.5
alpha = np.linspace(0.1, 1.0, N)*2e2
c = np.fft.fft(alpha) / N
C = circulant(c)
A = np.einsum("i, ij, j->ij", k, C, k)

# Show that A is Hermitian
print(np.allclose(A, A.conj().T))

# Arbitrary (random) initial vector
np.random.seed(0)
v = np.random.rand(N)
# Perform Arnoldi iteration with complex A
_, h = arnoldi_iteration(A, v, N)
# Perform Arnoldi iteration with real A
_, h2 = arnoldi_iteration(np.real(A), v, N)

# Plot results
plt.subplot(121)
plt.imshow(np.abs(h))
plt.title("Complex A")
plt.subplot(122)
plt.imshow(np.abs(h2))
plt.title("Real A")
plt.tight_layout()
plt.show()
```

True

```
<ipython-input-1-90a6c109bec1>:20: DeprecationWarning: `np.complex` is a deprecated alias for the builtin `complex`. To silence this warning, use `complex` by itself. Doing this will not modify any behavior and is safe. If you specifically wanted the numpy scalar type, use `np.complex128` here.
```

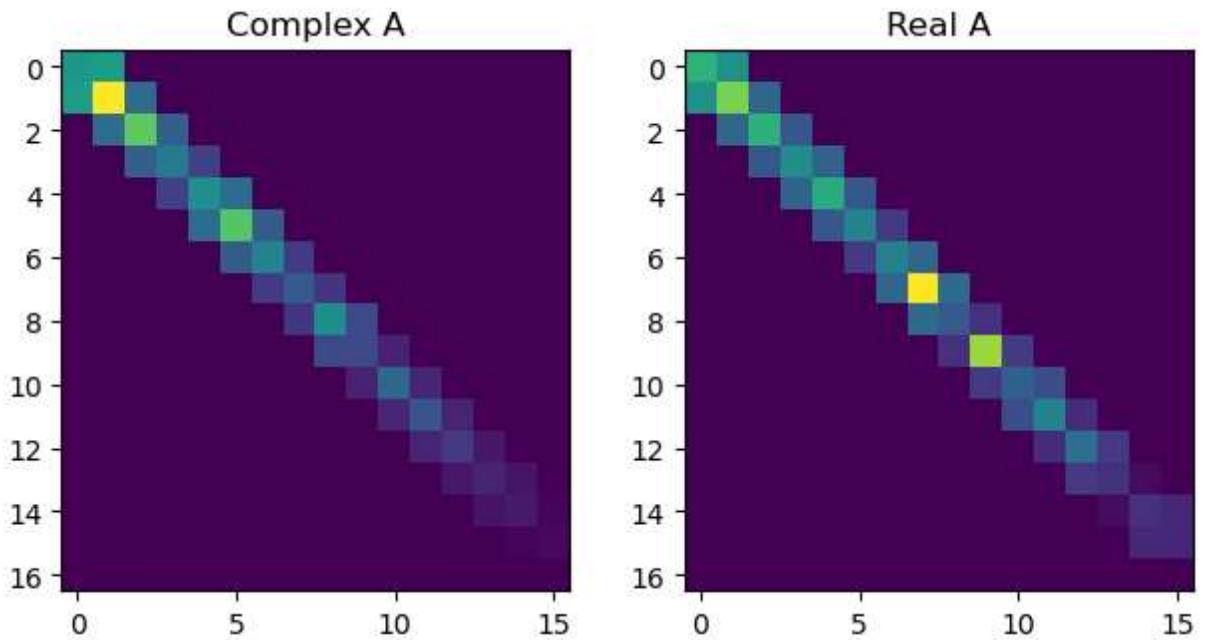
```
Deprecated in NumPy 1.20; for more details and guidance: https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations
```

```
.. h = np.zeros((n+1, n), dtype=np.complex)
```

```
<ipython-input-1-90a6c109bec1>:21: DeprecationWarning: `np.complex` is a deprecated alias for the builtin `complex`. To silence this warning, use `complex` by itself. Doing this will not modify any behavior and is safe. If you specifically wanted the numpy scalar type, use `np.complex128` here.
```

```
Deprecated in NumPy 1.20; for more details and guidance: https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations
```

```
.. Q = np.zeros((A.shape[0], n+1), dtype=np.complex)
```



In []:

```

..., h = lanczos_iteration(A, v, N)
# Perform Arnoldi iteration with real A
..., h2 = lanczos_iteration(np.real(A), v, N)

# Plot results
plt.subplot(121)
plt.imshow(np.abs(h))
plt.title("Complex A")
plt.subplot(122)
plt.imshow(np.abs(h2))
plt.title("Real A")
plt.tight_layout()
plt.show()

```

<ipython-input-2-6343734e5b98>:4: DeprecationWarning: `np.complex` is a deprecated alias for the builtin `complex`. To silence this warning, use `complex` by itself. Doing this will not modify any behavior and is safe. If you specifically wanted the numpy scalar type, use `np.complex128` here.

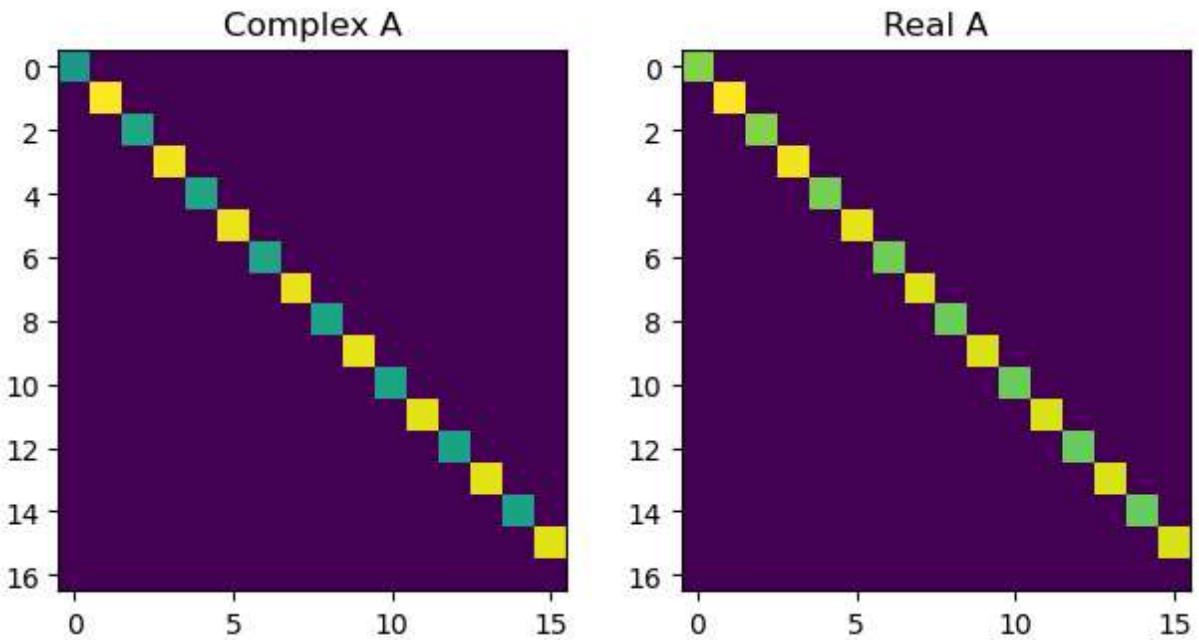
Deprecated in NumPy 1.20; for more details and guidance: <https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations>

... Q = np.zeros((A.shape[0], n+2), dtype=np.complex)

<ipython-input-2-6343734e5b98>:5: DeprecationWarning: `np.complex` is a deprecated alias for the builtin `complex`. To silence this warning, use `complex` by itself. Doing this will not modify any behavior and is safe. If you specifically wanted the numpy scalar type, use `np.complex128` here.

Deprecated in NumPy 1.20; for more details and guidance: <https://numpy.org/devdocs/release/1.20.0-notes.html#deprecations>

... h = np.zeros((n+1, n), dtype=np.complex)



QR iteration for Hessenburg matrix to find the eigenvalue decomposition.

```
In [ ]:
def Q_i(Q_min, i, j, k):
    if i < k or j < k:
        return float(i == j)
    else:
        return Q_min[i-k][j-k]

def QR_householder(A):
    n = A.shape[0]

    # Set R equal to A, and create Q as a zero matrix of the same size
    R = A.copy()
    Q = np.zeros((n, n))

    # The Householder procedure
    for k in range(n-1):
        I = np.eye(n)
        x = R[k:, k]
        e = I[k:, k]
        alpha = np.sign(x[0]) * np.linalg.norm(x)

        # Using anonymous functions, we create u and v
        u = np.array(list(map(lambda p, q: p + alpha * q, x, e)))
        print(u)
        norm_u = np.linalg.norm(u)
        v = np.array(list(map(lambda p: p/norm_u, u)))

        # Create the Q minor matrix
        Q_min = np.array([[float(i==j) - 2.0 * v[i] * v[j] for i in range(n-k)] for j in range(n-k+1)])
        Q_t = np.array([[Q_i(Q_min, i, j, k) for i in range(n)] for j in range(n)])

        if k == 0:
            Q = Q_t
            R = Q_t @ A
        else:
            Q = Q_t @ Q
```

```
R = Q_t@R
```

```
# Since Q is defined as the product of transposes of Q_t,  
# we need to take the transpose upon returning it  
return Q.T, R
```

T2

Computer problem (in C or C++): Write the functions to achieve Newton's method and Broyden's method. Test these methods for a few equation systems.

```
In [ ]:
```

```
from scipy import misc  
from scipy.optimize import newton  
  
# Newton's Method  
def NewtonsMethod(f, x, tolerance=0.000001):  
    while True:  
        x1 = x - f(x) / misc.derivative(f, x)  
        t = abs(x1 - x)  
        if t < tolerance:  
            break  
        x = x1  
    return x  
  
# Broyden's method  
def broyden(x, y, f_equations, J_equations, tol=10e-10, maxIters=50):  
    steps_taken = 0  
  
    f = f_equations(x, y)  
    J = J_equations(x, y)  
  
    while np.linalg.norm(f, 2) > tol and steps_taken < maxIters:  
        s = solve(J, -1*f)  
  
        x = x + s[0]  
        y = y + s[1]  
        newf = f_equations(x, y)  
        z = newf - f  
  
        J = J + (np.outer((z - np.dot(J, s)), s)) / (np.dot(s, s))  
        f = newf  
        steps_taken += 1  
  
    return steps_taken, x, y
```

```
In [ ]:
```

```
def f(x):  
    return (1.0/4.0)*x**3+(3.0/4.0)*x**2-(3.0/2.0)*x-2  
  
x = 4  
  
x_0 = NewtonsMethod(f, x)  
x_1 = newton(f, x, fprime=None, args=(), tol=1.48e-08, maxiter=50, fprime2=None)
```

```
In [ ]:
```

```
x_0, x_1
```

```
Out[ ]: (2.0000002745869883, 2.0000000000000008)
```

```
In [ ]:
```

```
tol = 10.0**-15
maxIters = 50
x0 = 1
y0 = 2

def fs(x, y):
    return np.array([x + 2*y - 2, x**2 + 4*y**2 - 4])

def Js(x, y):
    return np.array([[1, 2],
                   [2, 16]])

n, x, y = broyden(x0, y0, fs, Js, tol, maxIters=50)
print("iterations: ", n)
print("x and y: ", x, y)
```

```
iterations: 8
x and y: 8.159067075518965e-17 0.9999999999999999
```