

Linalg	=
LU	Skript S. 186
Cholesky	" S. 186
QR	S. 190
SVD	S. 196

## Ausgleichsrechnung (linear)

```
def solve-normal (A, b):
    A_H = A.conjugate().transpose()
    return solve(dot(A_H, A), dot(A_H, b))

def solve-gr (A, b):
    Q, R = scipy.linalg.qr(A, mode="economic")
    return solve-triangular(R, dot(Q.transpose(), b))
    scipy.linalg
```

Exact: `scipy.linalg.lstsq(A, b) [0]`

```
def solve-svd (A, b, eps=0):
    U, S, Vh = scipy.linalg.svd(A)
    r = 1 + where (S/S[0] > eps) [0].max()
    return dot(Vh[:, :r].T, dot(U[:, :r].T, b) / S[:, :r])
```

def solve-svd (A, b):

anz. messungen  
 $x = \text{linspace}(t_0, t_{end}, m)$   
 $y = \text{array mit messresultat}$

```
U, S, Vh = scipy.linalg.svd(A)
r = S.shape[0]; Ux = U[:, :r]
return dot(Vh[:, :r].T, dot(np.diag(1/S), dot(Ux.T, b)))
```

def monom-basis (deg):

```
basis = []
for n in range(deg+1):
    def monom(x, deg=n):
        return power(x, deg)
    basis.append(monom)
return basis
```

def build-system (x, y, basis):

```
n = len(basis); m = x.shape[0]
A = zeros((m, n))
for j in range(n):
    A[:, j] = basis[j](x)
return A, y
```

(nicht linear) S. 212

```
def Gauss-newton (F, DF, x0, tol=10e-6, maxit=100):
    for i in range(maxit):
        S = linalg.lstsq(DF(x0), F(x0)) [0]
        x0 = x0 - S
        if norm(S) < tol * norm(x0):
            return x0, True, i+1
    return x0, False, maxit
```

$$F = \begin{pmatrix} f(x, t_1) - y_1 \\ \vdots \\ f(x, t_m) - y_m \end{pmatrix}$$

DF = Jacobi Matrix von F

$$x = \text{argmin} \|F(x)\|_2^2$$

Residuenvektor  
 scipy.optimize.curve\_fit(F(x), x0)

Exact: `scipy.optimize.curve_fit(f, [t1, ..., tm], [y1, ..., ym]) [0]`

## Splitting

```
def compile_splitting_step(phi_a, phi_b, a, b):  
    def splitting_step(y0, dt):  
        y = y0  
        for i in range(a.shape[0]):  
            y = phi_a(y, a[i] * dt)  
            y = phi_b(y, b[i] * dt)  
        return y  
    return splitting_step
```

```
def integrate_splitting(step, y0, t0, tend, N)  
    b, dt = linspace(t0, tend, N, retstep=True)  
    y = zeros((b.shape[0], y0.shape[0]))  
    y[0] = y0  
    for i in range(N-1):  
        y[i+1] = step(y[i], dt)  
    return t, y
```

## Runge Kutta

class RKSolver:

```
def __init__(self, A, b, c):  
    s = A.shape[0]  
    if array_equal(tril(A, k=-1), A):  
        self.k = self.explicit_k  
    else:  
        self.k = self.implicit_k  
    self.__dict__.update(locals())
```

```
def explicit_k(self, k0, rhs, y0, t0, dt):  
    d = y0.shape[0]  
    k = zeros((self.s, d))  
    for i in range(self.s):  
        k[i] = rhs(y0 + dt * dot(self.A[i], k), t0 + dt * self.c[i])  
    return k
```

```
def k_equation(self, k, rhs, y0, t0, dt):  
    gk = zeros_like(k)  
    for i in range(self.s):  
        gk[i] = rhs(y0 + dt * dot(self.A[i], k), t0 + dt * self.c[i])  
    return k - gk
```

```
def implicit_k(self, k0, rhs, y0, t0, dt):  
    return fsolve2(self.k_equation, k0, args=(rhs, y0, t0, dt))
```

```
def integrate(self, rhs, y0, t0, tend, N):  
    d = y0.shape[0]; t, dt = linspace(t0, tend, N, retstep=True)  
    y = zeros((N, d)); y[0] = y0; k = zeros((self.s, d))  
    for n in range(N-1):  
        k = self.k(k, rhs, y[n], t[n], dt)  
        y[n+1] = y[n] + dt * dot(self.b, k)  
    return t, y
```

ausserhalb class

```
def fsolve2(f, x0, args):  
    shape = x0.shape  
    flat_x0 = x0.reshape((-1))  
    def flat_f(x, *args):  
        return f(x.reshape(shape), *args).reshape(-1)  
    x = fsolve(flat_f, flat_x0, args)  
    return x.reshape(shape)
```

Bsp

```
rk1 = RKSolver(A, b, c)  
t, y = rk1.integrate(rhs, y0, t0, tend, N)
```

ODE mit Matrix A

$$\dot{v} = A \cdot v$$

```
def integrak_krylov(A, v, dt, k, method):
```

```
    v, H = method(A, v, k)
```

```
    eH = scipy.linalg.expm(dt * H)
```

```
    r = dot(v[:, :-1], eH)
```

```
    return r[:, 0]
```

```
def integrak_scipy(A, v, dt):
```

```
    return dot(scipy.linalg.expm(dt * A), v)
```

```
def rkstep(f, y0, t0, h, s):
```

```
    k = zeros((y0.size, s))
```

```
    for i in range(s):
```

```
        k[:, i] = f(t0 + c[i] * h, y0 + h * dot(k, A[i]))
```

```
    y = y0 + h * dot(k, b)
```

```
    return y
```

```
def integrak(f, y0, t0, tend, N, A, b, c):
```

```
    T, h = linspace(t0, tend, N, retstep=True)
```

```
    Y = zeros((y0.size, N))
```

```
    Y[:, 0] = y0
```

```
    s = len(b)
```

```
    for i in range(N-1):
```

```
        Y[:, i+1] = rkstep(f, Y[:, i], T[i], h, s)
```

```
    return T, Y
```

```
def integrak_diagonalize(A, v, dt):
```

```
    eW, eV = scipy.linalg.eig(A)
```

```
    D = diag(exp(dt * eW))
```

```
    y = scipy.linalg.solve(eV, v)
```

```
    y = dot(D, y)
```

```
    y = dot(eV, y)
```

```
    return y
```

Rose-hrech-Euler

```
def expEV(N, t0, tend, y0, f, DF):
```

```
    ts, h = linspace(t0, tend, N, retstep=True)
```

```
    y0 = atleast_1d(y0)
```

```
    y = zeros((N, y0.shape[0]))
```

```
    y[0, :] = y0
```

```
    for i, ti in enumerate(ts[1:]):
```

```
        DF = atleast_2d(DF(y[i, :]))
```

```
        y[i+1, :] = y[i, :] + dot(expm(h * DF) - eye(*DF.shape),
```

```
                                solve(DF, f(y[i, :])))
```

```
    return ts, y
```

def Gr(a):

m, n = a.shape

q = zeros((m, n))

r = zeros((n, n))

for j in range(n):

vj = a[:, j].copy()

for i in range(j):

r[i, j] = np.dot(q[:, i], a[:, j])

vj = vj - r[i, j] \* q[:, i]

r[j, j] = norm(vj)

q[:, j] = vj / r[j, j]

return q, r

Phasenraum  $\Rightarrow$  q gegen p plotten

$$H = \frac{p^2}{2} + V(q)$$

$$\dot{q} = \frac{\partial H}{\partial p} = p$$

$$\dot{p} = -\frac{\partial H}{\partial q} = -\nabla V(q)$$

def Grmod(a):

m, n = a.shape

q = zeros((m, n))

r = zeros((n, n))

v = a.copy()

for i in range(n):

r[i, i] = norm(v[:, i])

q[:, i] = v[:, i] / r[i, i]

for j in range(i+1, n):

r[i, j] = dot(q[:, i], v[:, j])

v[:, j] = v[:, j] - r[i, j] \* q[:, i]

return q, r

$$Y = \begin{pmatrix} q \\ p \end{pmatrix}$$

$$\dot{Y} = \begin{pmatrix} \dot{q} \\ \dot{p} \end{pmatrix} = \begin{pmatrix} p \\ -\nabla V(q) \end{pmatrix} = \begin{pmatrix} p \\ 0 \end{pmatrix} + \begin{pmatrix} 0 \\ -\nabla V(q) \end{pmatrix}$$

Falls Reibung:

$$\ddot{q} + \mu \dot{q} + \nabla V(q) = 0$$

$$\dot{Y} = \begin{pmatrix} p \\ -\mu q - \nabla V(q) \end{pmatrix} = \begin{pmatrix} p \\ -\mu p - \nabla V(q) \end{pmatrix}$$

## Adaptives Verfahren

```
def integralk(f, y0, t0, tend, h0, hmax, atol):
```

```
    t = np.empty((1,))
```

```
    y = np.empty((1, y0.size))
```

```
    y[0] = y0
```

```
    t[0] = t0
```

```
    h = h0
```

```
    while (t[-1] != tend):
```

```
        y2 = method2(y0, f, h)
```

```
        y3 = method3(y0, f, h)
```

```
        if (norm(y2 - y3) < atol):
```

```
            y = np.vstack([y, y3])
```

```
            t = np.append(t, t[-1] + h)
```

```
            h = min([2 * h, tend - t[-1], hmax])
```

```
        else:
```

```
            h = 0.5 * h
```

```
    return t, y
```

## Methoden Stoermer-Verlet

```
def leap-frog(rhs, xy0, v0, W, tend):
```

```
    t, dt = linspace(0, tend, N+1, retstep=True)
```

```
    xy = zeros((N+1, xy0.size))
```

```
    v = zeros((N+1, v0.size))
```

```
    xy[0, :] = xy0
```

```
    v[0, :] = v0
```

```
    for i in range(N):
```

```
        v12 = v[i, :] + 0.5 * dt * rhs(xy[i, :])
```

```
        xy[i+1, :] = xy[i, :] + dt * v12
```

```
        v[i+1, :] = v12 + 0.5 * dt * rhs(xy[i+1, :])
```

```
    return t, xy
```

## 2D - Integration

```
F = lambda y: method(lambda x: f(x, y), a, b, Nx)
```

```
I = method(F, c, d, Ny)
```

```
in method f(x) mit array (list(map(f, x)))
```

```
ersetzten z.B.
```

```
sum(np.array(list(map(f, x[2:2:2])))
```

```
Diagonalisieren
```

```
skaliert sum(f(x[2:2:2]))
```

```
D, V = eig(A)
```

```
A = V @ diag(D) @ V-1
```

```
def semirk_step(y0, t0, h, f):
```

```
    S = A.shape[0]
```

```
    k = empty((y0.size, S))
```

```
    for i in range(S):
```

```
        * ->
```

```
        initial-guess = y0
```

```
        initial-guess = y0
```

```
        k[:, i] = fsolve(F, initial-guess)
```

```
        k[:, i] = fsolve(F, initial-guess)
```

```
        y1 = y0 + h * sum(k, axis=-1)
```

```
* F = lambda x: x - f(t + c[i] * h, y0 + h * sum(A[i, :i] * k[:, :i], axis=-1)
```

```
+ A[i, :i] * x)
```

```
def stoermer-verlet(rhs, xy0, v0, W, tend):
```

```
    t, dt = linspace(0, tend, N+1, retstep=True)
```

```
    xy = zeros((N+1, xy0.size))
```

```
    xy[0, :] = xy0
```

```
    xy[1, :] = xy[0, :] + dt * v0 + 0.5 * dt**2 * rhs(xy[0, :])
```

```
    for i in range(1, N):
```

```
        xy[i+1, :] = -xy[i-1, :] + 2 * xy[i, :] + dt**2 * rhs(xy[i, :])
```

```
    return t, xy
```

```
def lemand-acceleration(xy):
```

```
    r = linlg.norm(xy)
```

```
    dudt = 2 * xy * (2 / r**14 - 1 / r**8)
```

```
    return dudt
```

```

def integrate(method, rhs, y0, T, N):
    # array [x, v]
    y = empty((N+1) * y0.shape)
    t0, dt = 0, T/N
    y[0, ...] = y0
    for i in range(0, N):
        y[i+1, ...] = method(rhs, y[i, ...], t0+i*dt, dt)
    t = arange(N+1) * dt
    return t, y

```

Newton ohne lstsq.

```

def newton(f, df, x0, L=1, rtol=1e-14, maxit=1000)

```

x = x0

for i in range(maxit)

delta = L \* solve(df(x), f(x))

x -= delta

if norm(delta) < rtol

return x, True, i+1

return x, False, maxit

```

def velocity_velt_step(rhs, xv0, t0, dt):
    xv0 = xv0.reshape((2, -1))
    xvL = np.empty_like(xv0)
    x0, xL = xv0[0, :], xvL[0, :]
    v0, vL = xv0[1, :], xvL[1, :]
    xL[:] = x0 + dt * v0 + 0.5 * dt**2 * rhs(t0, x0)
    vL[:] = v0 + 0.5 * dt * (rhs(t0, x0) + rhs(t0+dt, xL))
    return xvL.reshape(-1)

```

```

def velocity_velt(rhs, y0, T, N):

```

return integrate(velocity\_velt\_step, rhs, y0, T, N)

$$R(p) = \begin{pmatrix} F(t_1, p) - y_1 \\ \vdots \\ F(t_n, p) - y_n \end{pmatrix}$$

df ist nach p[0], p[1], ..., p[k]

scipy.optimize.leastsq(R(x), x0)

# Quadratur

```

Exact: scipy.integrate.quad(f,a,b)[0]
x, h = linspace(a,b,N+1,retstep=True)
L = x + 0.5h
return sum(h*f(c[:1]))

```

golub/wetzel: Skript S. 17

Gaussquad: Skript S. 19

Adapt. quad: Skript S. 23

composite-legendre, Zusammenfassung der Gauss-legendre

```

def composite-legendre-error(f, exact)
n = ...
rule = lambda f, a, b, N: comp.legendre(f, a, b, N, n)
errors, n_chunks = quaderror(rule, f, exact)
return errors, n_chunks

```

## def MPR(f, a, b, N)

```

x, h = linspace(a,b,N+1,retstep=True)
L = x + 0.5h
return sum(h*f(c[:1]))

```

## order

```

N_array = 2**arange(5,12)
error = zeros_like(N_array, dtype=float)
I_exact = integrate.quad(f, a, b)[0]
for j, N in enumerate(N_array):
    I = method(f, a, b, N)
    error[j] = abs(I - I_exact)
order = polyfit(log(N_array), log(error), 1)[0]
plt.loglog(N_array, error)

```

anzahl Quadratpunkte

```

def gauss-legendre(f, a, b, n)
x_ref, w_ref = golub_wetzel(n)
x = a + (x_ref + 1.0) * (b-a) * 0.5
w = 0.5 * (b-a) * w_ref
return np.sum(w * f(x))

```

anz. Teilintervalle

## def composite-legendre(f, a, b, N, n)

```

I = 0
dx = (b-a)/N
for i in range(N):
    I += gauss-legendre(f, a + i*dx, a + (i+1)*dx, n)
return I

```

# DGL

f(y, t)

```

Exact: def scipy(f, y0, t0, tend, N):
T = linspace(t0, tend, N)
return T, odeint(f, y0, T)

```

Engie

Sci.integr.

```

def eE/iE/IMP(f, y0, t0, tend, N):
t, dt = linspace(t0, tend, N, retstep=True)
y = zeros((t.shape[0], y0.shape[0]))
y[0] = y0
for k in range(t.shape[0]-1):

```

```

(ee) y[k+1] = y[k] + dt * f(y[k], t[k])

```

```

(ie) y[k+1] = fsolve(lambda yk1: y[k] + dt * f(yk1, t[k+1]) - yk1, y[k])

```

```

(imp) y[k+1] = fsolve(lambda yk1: y[k] + dt * f(0.5*(y[k] + yk1), 0.5*(t[k] + t[k+1])) - yk1, y[k])

```

return t, y

## order

```

N_array = 2**arange(5,12)
errors = zeros_like(N_array, dtype=float)
for j, N in enumerate(N_array):
    t, y = method(-, N)
    errors[j] = linAlg.norm(y[-1] - y_exact[-1])
    errors[j] = norm(y - y_exact)
    errors[j] = norm(y - y_exact, axis=1).max()
order = polyfit(log(N_array), log(errors), 1)[0]

```

Stilles resultat matric

allgemein

max. norm

```

def Stoermer-verlet(f, y0, v0, t0, tend, N):
t, dt = linspace(t0, tend, N, retstep=True)
y = zeros(N)
y[0] = y0
y[1] = y0 + dt * v0 + dt**2 * 0.5 * f(y0, t0)
for k in range(1, N-1):
    y[k+1] = -y[k-1] + 2 * y[k] + dt**2 * f(y[k], t[k])
return t, y

```

## def velocity-verlet(f, y0, v0, t0, T, N):

```

t, dt = ...
y = zeros(N); y[0] = y0
v = zeros(N); v[0] = v0
for k in range(N-1):
    y[k+1] = y[k] + dt * v[k] + dt**2 * 0.5 * f(y[k], t[k])
    v[k+1] = v[k] + dt * 0.5 * f(y[k], t[k]) + f(y[k+1], t[k+1])
return t, y, v

```

Engie

```

t, y = ode45(f, [t0, tend], y0)

```

f(t, z) array

Eigenwerte eigenwert, vektoren eig falls hermitesch/symmetrisch

Exact:  $W, V = \text{scipy.linalg.eig}(A)$

Algorithmen S. 223

S. 230

def inverse-potenz-methode(A, rtol=10e-6):

LUP = scipy.linalg.lu\_factor(A)

x = random.rand(A.shape[0])

x /= np.linalg.norm(x)

ew\_before = 0

while (True):

    Lu\_solve(LUP, x, overwrite\_b=True)

    ew = 1 / norm(x)

    x\* = ew

    if abs(ew - ew\_before) < rtol \* ew:  
        break

    ew\_before = ew

return np.dot(x, transpose(x)), np.dot(A, x)

S. 237

def arnoldi(A, v0, k):

r, c = A.shape

V = zeros((r, k+1), dtype=complexfloating)

H = zeros((k+1, k), dtype=complexfloating)

V[:, 0] = v0.reshape(-1) / norm(v0)

for i in range(1, k+1):

    Vi = dot(A, V[:, i-1])

    for j in range(i):

        H[j, i-1] = dot(conjugate(V[:, j]), Vi)

        Vi -= H[j, i-1] \* V[:, j]

    H[i, i-1] = norm(Vi)

    V[:, i] = Vi / H[i, i-1]

return V, H[0:k, :]

ew = np.linalg.eigvals(H[0:k, :])

ew, ev = eig(H[0:k, :]) I = argsort(ew) ev = dot(V[:, I-1], ev[:, I])

def potenz-methode(A, rtol=10e-6):

x = np.random.rand(A.shape[0])

x /= np.linalg.norm(x)

ew = 1 ; ew\_before = 0

while (abs(ew - ew\_before) > rtol \* ew):

    ew\_before = ew

    x = dot(A, x)

    ew = linalg.norm(x)

    x /= ew

return ew

def shifted-inverse(A, c, rtol=10e-6):

return inverse-potenz-methode(A - c \* np.eye(A.shape[0])) + c

S. 239

def lanczos(A, v0, k):

r, c = A.shape

V = zeros((r, k+1), dtype=complexfloating)

H = zeros((k+1, k), dtype=complexfloating)

alpha = zeros((k+1, ), dtype=complexfloating)

beta = zeros((k+1, ), dtype=complexfloating)

V[:, 0] = v0.reshape(-1) / norm(v0)

for i in range(1, k+1):

    Vi = dot(A, V[:, i-1])

    if i > 1:

        Vi -= beta[i-1] \* V[:, i-2]

    alpha[i-1] = dot(conjugate(V[:, i-1]), Vi)

    Vi -= alpha[i-1] \* V[:, i-1]

    beta[i] = norm(Vi)

    V[:, i] = Vi / beta[i]

beta = beta[1:-1]

H = diag(alpha) + diag(beta, 1) + diag(beta, -1)

return V, H

ew = np.linalg.eigvals(H[0:k, :])



# Nullstellen

- 10) Bisektion S. 140
- Sekanten S. 143

```
def Newton_10(f, df, x0, L=1.0, rtol=1e-14, maxit=1000):
    x = x0
    for i in range(maxit):
        delta = L * solve(df(x), f(x))
        x -= delta
        if norm(delta) < rtol:
            return x, True, i+1
    return x, False, maxit
```

## def niger.fsolve(F, guess)

```
shape = guess.shape
guess = guess.reshape((-1,))
result = fsolve(lambda x: F(x.reshape(shape)).reshape((-1,)), guess)
return result.reshape(shape)
```

## def Householder (A)

```
L = len(A)
for i in range(0, L-1):
    I = np.eye(L-1)
    ai = A[:, i]
    abs = norm(ai[1:L])
    sign = np.sign(A[1, i])
    vi = array([ai[1:L]]).T + sign * abs * array([I[:, 0]]).T
    Qi = I - 2 * (vi @ vi.T) / (vi.T @ vi)
    H = np.eye(L)
    H[i:L, i:L] = Qi
    R = J @ A
return R, A, H
```

## Householder

$$H(v) = I - 2 \frac{v \cdot v^H}{v^H \cdot v}$$

$$v = \begin{cases} \frac{1}{2} (a \pm \|a\|_2 e_1), & a_1 > 0 \\ \frac{1}{2} (a - \|a\|_2 e_1), & a_1 < 0 \end{cases}$$

## Givens Rotation

$$Q \cdot A = R$$

$$Q = \begin{pmatrix} \cos \varphi & -\sin \varphi \\ \sin \varphi & \cos \varphi \end{pmatrix}$$

## Lösungsschritte

$$C = \frac{a_{11}}{\sqrt{a_{21}^2 + a_{11}^2}}$$

$$S = \frac{a_{21}}{\sqrt{a_{21}^2 + a_{11}^2}}$$

$$Q_1 = \begin{pmatrix} C & S & 0 \\ -S & C & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

→ dieser Eintrag wird zu 0 1)

$$Q_1 \cdot A = B$$

2.) Bearbeite B mit  $Q_2$

3.)  $C = Q_2 \cdot B$

4.) Bearbeite C mit  $Q_3$

5.)  $Q = Q_3 \cdot Q_2 \cdot Q_1$

Exact: scipy.optimize.fsolve(f, y0)  
↳  $\mathbb{R}^n \rightarrow \mathbb{R}^n$

## methoden

- Newton S. 145 altest 2d löschen
- dampNewton S. 150
- Broyden S. 154 ( $J = Df(y_0)$ )

konv Ordnung  $p \approx \frac{\log(e^{k+1}) - \log e^k}{\log e^k - \log e^{k-1}}$

$e^k =$  fehler bei k

Code:

```
p = log(e[L:]) / e[1:-1] / log(e[1:-1] / e[1:-2])
paverage = sum(p) / size(p) + 1e-21
plt.semilogy(k, e)
fit = polyfit(k, log(e), 1)
L = exp(fit[0])
```

## Givens Rotation