

CP1 Blatt4 Abgabe Lapp & Brieden

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Aufgabe 4.1 Prandtl-van Karman law ¶

Function for friction factor: $f = g(f) = \frac{1}{(0.88 * \ln(Re\sqrt{f}) - 0.81)^2}$

the derivative: $g'(f) = \frac{-1}{\pm f * (0.88 * \ln(Re\sqrt{f}) - 0.81)^3}$

In [2]:

```
def factor_f(f0, re, acc):
    df = 100000      #Initailisierung für while()-Bed
    dif_gf=0

    while (dif_gf<1 and df>acc):
        f_np1 = 1/((0.88*math.log(re*math.sqrt(f0))-0.81)**2) # f=g(f)

        dif_gf= math.fabs(1/(f_np1*(0.88*math.log(re*math.sqrt(f_np1))-0.81)**3))#f=g'(f)
        df=math.fabs(f_np1-f0)      #Check Abbruch

        f0=f_np1                    # Wert zuweisen für nächste Iteration
        #print(f_np1,dif_gf) #Check Werte
    return f_np1

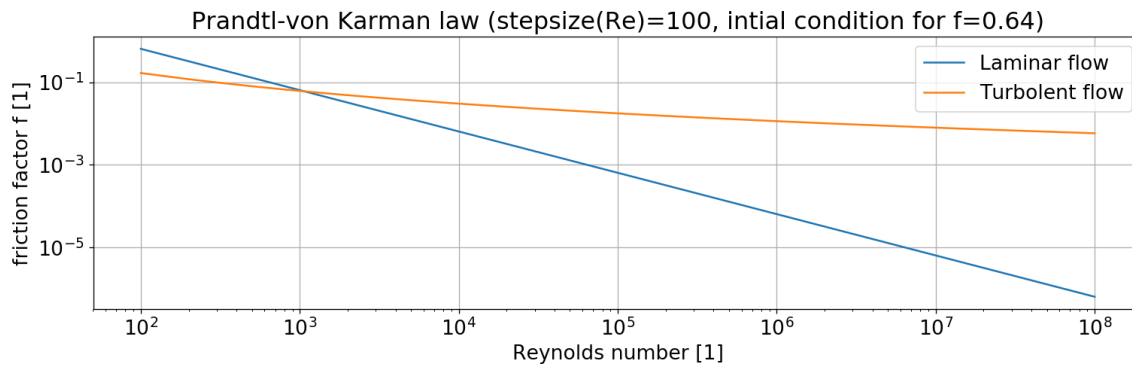
acc = 0.00001 #Genauigkeit f
step = 100    #Schrittweite
re_dat= np.arange(10**2,10**8,step)

f0_turb = 0.64 #Anfangswert f
#print(factor_f(f0_turb,100,acc))
values_turb = np.array([factor_f(f0_turb,re_,acc)for re_ in re_dat])
```

In [3]:

```
plt.loglog(re_dat,64/re_dat,label="Laminar flow")
plt.loglog(re_dat,values_turb,label="Turbulent flow")

plt.xlabel('Reynolds number [1]')
plt.ylabel('friction factor f [1]')
plt.legend(loc='upper right')
plt.title('Prandtl-von Karman law (stepsize(Re)=100, intial condition for f=0.64)')
plt.grid(True)
plt.show()
```



Another way of determining the curve $f(Re)$ is to use the Colebrook equation:

$$\frac{1}{\sqrt{f}} = -2 * \log\left(\frac{2.51}{Re * \sqrt{f}} + \frac{k}{3.72d_h}\right)$$

d_h : Hydraulic diameter

k : Roughness of the pipe

Re : Reynold number

Aufgabe 4.2. Lagrange points

$$H = \frac{p_x^2 + p_y^2}{2} - xp_y + yp_x - \frac{m_1}{\sqrt{(x+m_2)^2 + y^2}} - \frac{m_2}{\sqrt{(x-m_1)^2 + y^2}}$$

is composite with the following parts:

- with force on particle i: $\vec{F}_i = -G \sum_{j=1}^N \frac{m_i m_j}{|\vec{r}_j - \vec{r}_i|^3} (\vec{r}_j - \vec{r}_i) = -\vec{\nabla}_{\vec{r}_i} V$
- the following potential appears on particle i:

$$V = -G \sum_{i < j} \frac{m_i m_j}{|\vec{r}_j - \vec{r}_i|} = -G \left(\frac{m_i m_1}{|\vec{r}_1 - \vec{r}_i|} + \frac{m_i m_2}{|\vec{r}_2 - \vec{r}_i|} \right)$$
- $\frac{p^2}{2m}$ or in our case with special units $\frac{p_x^2 + p_y^2}{2}$ describes the kinetic energy

Aufgabe 3.3. Residuals

In [4]:

```
A = np.array([[1.2969, 0.8648],[0.2161, 0.1441]])
x = np.array([0.9911,-0.4870])
b = np.array([0.8642,0.1440])
x_calculated = np.linalg.lstsq(A,b)[0]
print(" residual with given values :", np.dot(A,x.T)-b,
      "\n residual with calculated x :", np.dot(A,x_calculated.T)-b,
      "\n with LU-decomposition we get the x:", x_calculated)
```

```
residual with given values : [ -1.00000002e-08  1.00000000e-08]
residual with calculated x : [ -1.11022302e-16  2.77555756e-17]
with LU-decomposition we get the x: [ 2. -2.]
```

Beide Residuen sind sehr klein, aber das Residuum mit berechnetem x-Vektor ist nochmals 8 Größenordnungen kleiner. Das neu berechnete x ist genauer und liegt bei der Maschinengenauigkeit.

Aufgabe 4.4: Powers of Integers

In [5]:

```
# LGS der Form: Ax=b
# (Exponent,Basis)
def vektor_b(k,N): # gibt die einzelnen akkumulierten Teilsummen aus, für LGS als Vektor b
    n=1
    erg_sigma=[[0 for el in range(1) ]for zeilen in range(0,N,1)]
    sigma=0
    while n<=N:
        summand = n**k
        sigma=sigma+summand
        erg_sigma[n-1][0]=sigma
        n=n+1
    return erg_sigma # Elemente von der Liste erg_sigma über erg_sigma[0,1,2...] anspreche
n

#print(sum(2,3))

def matrix_A(k,N):

    mat_a=[[0 for zeilen in range(0,N,1)]for spalten in range(0,k+1,1)]

    for n in range(1,N+1,1):
        for l in range(1,k+2,1):
            pot= n**l
            mat_a[n-1][l-1]=pot # Auf Dimension achten, da solver sonst falsch berechnet
                                # ACHTUNG darstellung als Liste!!!

    return mat_a

# Für jedes k eine zeile von koeffizienten

k_lauf=7 # Berechnung für Potenzen bis k

for k in range(1,k_lauf+1,1):
    N=k+1
    print('\n For k = %i'%k)
    vector_ck =solve(matrix_A(k,N),vektor_b(k,N))

    r= np.dot(matrix_A(k,N),vector_ck)-vektor_b(k,N)
    print("Residual r = \n",r)
    print('Vector of c1 to ck+1 = \n',vector_ck)

    cond = np.linalg.cond(matrix_A(k,N),p=2) #np.linalg.con(A,p=...) Kondition einer Matrix
                                              # p=2 für 2-er Norm -> largest singular Value
    print('Condition of the Matrix A =\n %f'%cond)
```

```
For k = 1
Residual r =
[[ 0.]
 [ 0.]]
Vector of c1 to ck+1 =
[[ 0.5]
 [ 0.5]]
Condition of the Matrix A =
10.908327
```

```
For k = 2
Residual r =
[[ 4.44089210e-16]
 [ 0.00000000e+00]
 [ 0.00000000e+00]]
Vector of c1 to ck+1 =
[[ 0.16666667]
 [ 0.5       ]
 [ 0.33333333]]
Condition of the Matrix A =
141.235621
```

```
For k = 3
Residual r =
[[ 0.]
 [ 0.]
 [ 0.]
 [ 0.]]
Vector of c1 to ck+1 =
[[ 0.  ]
 [ 0.25]
 [ 0.5 ]
 [ 0.25]]
Condition of the Matrix A =
2501.238500
```

```
For k = 4
Residual r =
[[ -4.34097203e-14]
 [ -1.77635684e-14]
 [ 0.00000000e+00]
 [ -5.68434189e-14]
 [ 0.00000000e+00]]
Vector of c1 to ck+1 =
[[ -3.33333333e-02]
 [ 2.30926389e-14]
 [ 3.33333333e-01]
 [ 5.00000000e-01]
 [ 2.00000000e-01]]
Condition of the Matrix A =
56895.788056
```

```
For k = 5
Residual r =
[[ -1.30340183e-13]
 [ -4.68958206e-13]
 [ 0.00000000e+00]
 [ -2.27373675e-13]
 [ -9.09494702e-13]
 [ 0.00000000e+00]]
Vector of c1 to ck+1 =
```

```
[ [ 2.40918396e-12]
[ -8.33333333e-02]
[ 3.28128635e-12]
[ 4.16666667e-01]
[ 5.00000000e-01]
[ 1.66666667e-01]]
Condition of the Matrix A =
1589237.416884
```

```
For k = 6
Residual r =
[ [ -6.60027588e-13]
[ -9.52127266e-13]
[ 6.36646291e-12]
[ 5.45696821e-12]
[ -2.18278728e-11]
[ 1.45519152e-11]
[ 0.00000000e+00]]
Vector of c1 to ck+1 =
[ [ 2.38095237e-02]
[ 1.88592253e-10]
[ -1.66666667e-01]
[ 8.21778201e-11]
[ 5.00000000e-01]
[ 5.00000000e-01]
[ 1.42857143e-01]]
Condition of the Matrix A =
52843558.730114
```

```
For k = 7
Residual r =
[ [ -1.72173387e-12]
[ 1.12549969e-11]
[ 1.40971679e-11]
[ 1.09139364e-11]
[ -1.45519152e-11]
[ -1.16415322e-10]
[ 0.00000000e+00]
[ 0.00000000e+00]]
Vector of c1 to ck+1 =
[ [ 1.30010336e-10]
[ 8.33333330e-02]
[ 4.01905472e-10]
[ -2.91666667e-01]
[ 6.26576886e-11]
[ 5.83333333e-01]
[ 5.00000000e-01]
[ 1.25000000e-01]]
Condition of the Matrix A =
2042493005.718225
```