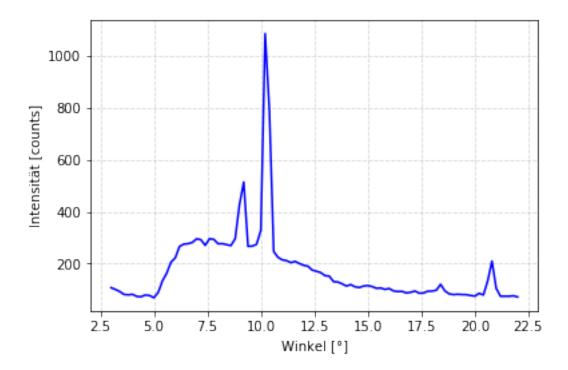
### 255

### May 19, 2024

```
[1]: %matplotlib inline
import numpy as np
import matplotlib.pyplot as plt
import matplotlib.mlab as mlab
from scipy.optimize import curve_fit
from scipy.stats import chi2
from scipy.stats import norm
import scipy.constants as scp
from scipy.integrate import quad
from tabulate import tabulate
from scipy import signal
import scipy.constants as const
[2]: def comma_to_float(valstr):
    return float(valstr.decode("utf-8").replace(',', '.'))
```

# 1 Zu Aufgabe 1

### 1.1 Grenzwellenlänge & Planck'sches Wirkungsqant



```
[5]: #background:
bkg = np.mean(inten_1[:10])
dbkg = np.std(inten_1[:10], ddof=1)/np.sqrt(10)
print('Hintergrund = ({} +/- {})counts'.format(bkg, dbkg))
inten_1 = inten_1 - bkg
```

Hintergrund = (84.98 + / - 3.699663648375499) counts

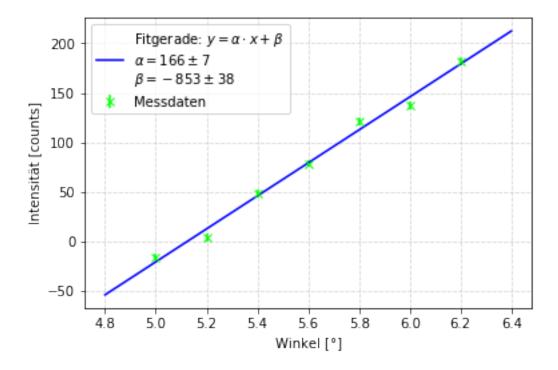
```
[6]: # beschränkung auf bereich zwischen 5 und 6.2 - Werte 11 bis 17
angle_1_lin = angle_1[10:17]
inten_1_lin = inten_1[10:17]

#stat. fehler:
dinten = np.sqrt((0.01 * inten_1_lin)**2 + dbkg**2)

def linfit(x, a,b):
    return a*x+b

popt_1, pcov_1 = curve_fit(linfit, angle_1_lin, inten_1_lin, sigma=dinten)
```

```
[7]: plt.grid(alpha=0.5, linestyle='--')
plt.xlabel('Winkel [°]')
plt.ylabel('Intensität [counts]')
```



```
Grenzwinkel = (5.124092937715125 +/- 0.31012283079239883)° Grenzwellenlänge = (3.5975326890255045e-11 +/- 2.1715082114079383e-12)m
```

```
[9]: #planck:
h_1 = lamb_gr_35 * const.e * 35000 /const.c
dh_1 = dlamb_gr_35 * const.e * 35000 /const.c
print("Planck'sches Wirkungsquant h = ({} +/- {})Js".format(h_1, dh_1))
print('Literaturwert: h = {} Js'.format(const.h))
```

Planck'sches Wirkungsquant h = (6.729185245356466e-34 +/-4.0618063210246797e-35) Js Literaturwert: h = 6.62607015e-34 Js

```
[10]: #Signifikanztest:
sign_h_1 = np.abs(h_1 - const.h)/dh_1
print('sigma_h =', sign_h_1)
```

 $sigma_h = 0.2538651210982732$ 

#### 1.2 Startwinkel des Spektrums 2. Ordnung

Startwinkel 2. Ordnung = (10.28975301540787 +/- 0.6278648607474591)°

# 2 Zu Aufgabe 2

#### 2.1 Wellenlänge der K-Linien 1. und 2. Ordnung

```
[12]: angle_2a , inten_2a = np.loadtxt('./TeamChopper_2.txt', unpack=True, unconverters={0:comma_to_float, 1:comma_to_float})

angle_2b , inten_2b = np.loadtxt('./TeamChopper_2b.txt', unpack=True, un
```

```
[13]: peaks_2a, _ = signal.find_peaks(inten_2a, height=400)
peaks_2b, _ = signal.find_peaks(inten_2b, height=100)
```

```
r'P1: ({}, {})'.format(angle_2a[peaks_2a[0]],⊔

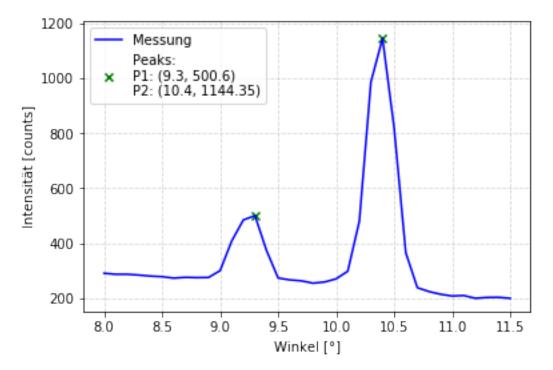
inten_2a[peaks_2a[0]]),

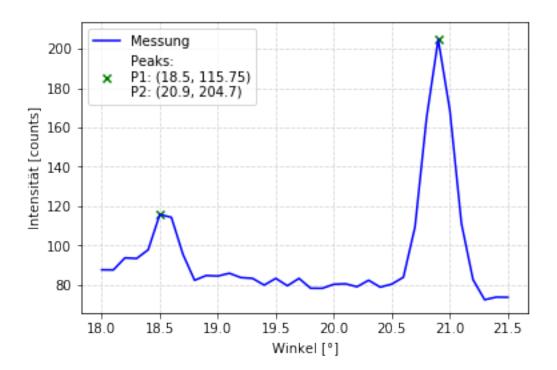
r'P2: ({}, {})'.format(angle_2a[peaks_2a[1]],⊔

inten_2a[peaks_2a[1]])]))

= plt.legend()

plt.savefig('./plots/A2.1.pdf', format='PDF')
```





```
beta_ka1 = angle_2a[peaks_2a[1]]
      beta_kb2 = angle_2b[peaks_2b[0]]
      beta_ka2 = angle_2b[peaks_2b[1]]
      #abschätzung aus abstand zwischen den aufgenommenen Werten:
      dbeta = 0.1
[17]: #Wellenlängen:
      lamb_ka1 = 2 * d * np.sin(np.radians(beta_ka1))
      lamb_kb1 = 2 * d * np.sin(np.radians(beta_kb1))
      lamb_ka2 = d * np.sin(np.radians(beta_ka2))
      lamb kb2 = d * np.sin(np.radians(beta kb2))
      dlamb_ka1 = 2 * d * np.cos(np.radians(beta_ka1)) * np.radians(dbeta)
      dlamb_kb1 = 2 * d * np.cos(np.radians(beta_kb1)) * np.radians(dbeta)
      dlamb_ka2 = d * np.cos(np.radians(beta_ka2)) * np.radians(dbeta)
      dlamb_kb2 = d * np.cos(np.radians(beta_kb2)) * np.radians(dbeta)
      print('K_a1 = ({} +/- {})m'.format(lamb_ka1, dlamb_ka1))
      print("K_b1 = ({} +/- {})m".format(lamb_kb1, dlamb_kb1))
      print("K_a2 = ({} +/- {})m".format(lamb_ka2, dlamb_ka2))
      print("K_b2 = ({} +/- {})m".format(lamb_kb2, dlamb_kb2))
```

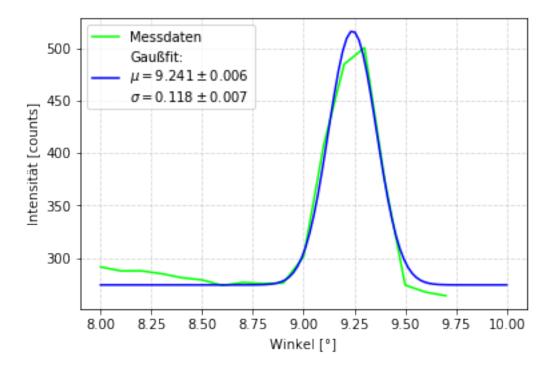
[16]: beta\_kb1 = angle\_2a[peaks\_2a[0]]

 $K_a1 = (7.271311170692555e-11 +/- 6.914690607415013e-13)m$  $K_b1 = (6.509401914043387e-11 +/- 6.937779465552078e-13)m$ 

```
K_a2 = (7.18470330629725e-11 +/- 3.283815714302227e-13)m
     K_b2 = (6.390515779998556e-11 +/- 3.333445949800182e-13)m
[18]: ka lit = 71.1e-12
     kb lit = 63.2e-12
      sigma_ka1 = np.abs(lamb_ka1 - ka_lit)/np.sqrt(dlamb_ka1**2)
      sigma_ka2 = np.abs(lamb_ka2 - ka_lit)/np.sqrt(dlamb_ka2**2)
      sigma_kb1 = np.abs(lamb_kb1 - kb_lit)/np.sqrt(dlamb_kb1**2)
      sigma_kb2 = np.abs(lamb_kb2 - kb_lit)/np.sqrt(dlamb_kb2**2)
      print(sigma_ka1, sigma_ka2)
      print(sigma_kb1, sigma_kb2)
     2.332876188553856 2.2748933800361892
     2.730007706122946 2.115401931229231
[19]: #tabelle:
      ordn = np.array([1,1,2,2])
      ab = np.array(['a','b','a','b'])
      angles = np.array([beta_ka1, beta_kb1, beta_ka2, beta_kb2])
      dangles = np.full(4, 0.1)
      lambs = np.array([lamb_ka1, lamb_kb1, lamb_ka2, lamb_kb2])
      dlambs = np.array([dlamb_ka1, dlamb_kb1, dlamb_ka2, dlamb_kb2])
      sigs = np.array([sigma_ka1, sigma_kb1, sigma_ka2, sigma_kb2])
     head1 = ['Ordn', 'a/b', 'angle', 'da', 'lamb', 'dl', 'sig']
      tab1 = zip(ordn, ab, angles, dangles, lambs, dlambs, sigs)
      print(tabulate(tab1, headers=head1, tablefmt="latex"))
     \begin{tabular}{rlrrrrr}
     \hline
        Ordn & a/b & angle & da &
                                               lamb &
                                                              dl &
                                                                        sig \\
     \hline
                     & 10.4 & 0.1 & 7.27131e-11 & 6.91469e-13 & 2.33288 \\
           1 & a
           1 & b
                         9.3 & 0.1 & 6.5094e-11 & 6.93778e-13 & 2.73001 \\
                    & 20.9 & 0.1 & 7.1847e-11 & 3.28382e-13 & 2.27489 \\
           2 & a
           2 & b
                   & 18.5 & 0.1 & 6.39052e-11 & 3.33345e-13 & 2.1154 \\
     \hline
     \end{tabular}
     2.2 Bestimmung der Halbwertsbreite für die Linien 1. Ordnung
[20]: def gaussian(x, mu, sig, a, bkg): #bkg = background
         return a/(sig * np.sqrt(2 * np.pi)) * np.exp(-0.5 * ((x - mu)/sig)**2) + bkg
[21]: popt_ka1, pcov_ka1 = curve_fit(gaussian, angle_2a[:20], inten_2a[:20], p0=[9.
      \rightarrow 5, 0.2, 500, 220])
```

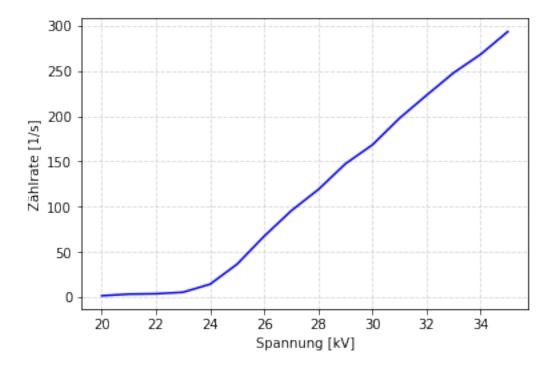
```
print(popt_ka1)
```

#### [9.24062762e+00 1.18287079e-01 7.19091385e+01 2.74097931e+02]



# 3 Zu Aufgabe 3

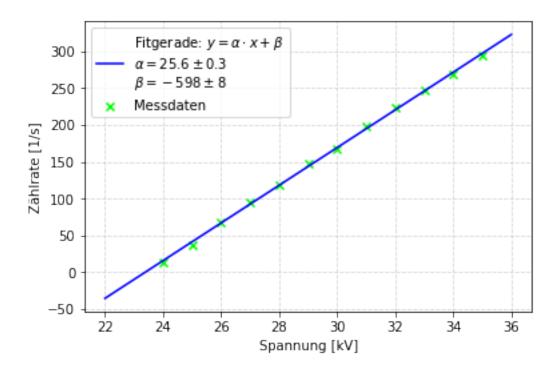
```
[24]: plt.grid(alpha=0.5, linestyle='--')
  plt.xlabel('Spannung [kV]')
  plt.ylabel('Zählrate [1/s]')
  plt.plot(U,rate, color='blue')
  plt.savefig('./plots/A3.1.pdf', format='PDF')
```



```
[25]: #extrapolation der linearen werte:
      U_lin = U[4:]
      rate_lin = rate[4:]
      popt_3, pcov_3 = curve_fit(linfit, U_lin, rate_lin)
      plt.grid(alpha=0.5, linestyle='--')
      plt.xlabel('Spannung [kV]')
      plt.ylabel('Zählrate [1/s]')
      plt.scatter(U_lin, rate_lin, color='lime', marker='x', label='Messdaten')
      x = np.linspace(22,36,100)
      plt.plot(x, linfit(x, *popt_3), color='blue',
              label="\n".join([r"Fitgerade: $y = \alpha \cdot x + \beta$",
                                 r'$\alpha ={:.1f}\pm{:.1}$'.format(popt_3[0], np.
       \rightarrowsqrt(pcov_3[0][0])),
                                 r'$\beta =\{:.0f}\pm\{:.0f}\$'.format(popt_3[1], np.

sqrt(pcov_3[1][1]))]))
      _ = plt.legend()
```

```
plt.savefig('./plots/A3fit.pdf', format='PDF')
```



Grenzspannung = (23.376873398337334 +/- 0.3778752487336699)kV

```
[27]: beta = 7.5 #°

#planck:
h_2 = 2 * d * np.sin(np.radians(beta)) * const.e * U_gr * 1000 /const.c
dh_2 = 2 * d * np.sin(np.radians(beta)) * const.e * dU_gr * 1000 /const.c
print("Planck'sches Wirkungsquant h = ({} +/- {})Js".format(h_2, dh_2))
print('Literaturwert: h = {} Js'.format(const.h))
```

Planck'sches Wirkungsquant h = (6.568455262011514e-34 +/-1.0617573289784417e-35) Js Literaturwert: h = 6.62607015e-34 Js

```
[28]: #Signifikanztest:
sign_h_2 = np.abs(h_2 - const.h)/dh_2
```

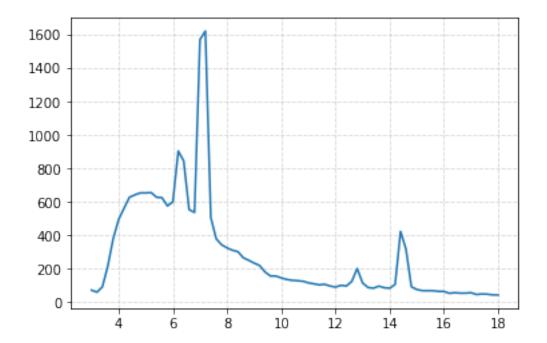
```
print('sigma_h =', sign_h_2)
```

 $sigma_h = 0.5426370641954453$ 

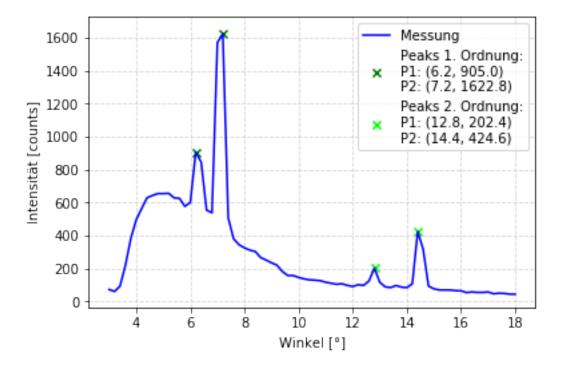
# 4 Zu Aufgabe 4

### 4.1 Position der K-Linien

```
[30]: plt.plot(angle_4, inten_4) plt.grid(alpha=0.5, linestyle='--')
```



```
[31]: peaks_4a, _ = signal.find_peaks(inten_4, height=700)
peaks_4b, _ = signal.find_peaks(inten_4[46:], height=150)
```



## 4.2 Gitterkonstante und Avogadro-Zahl

```
[33]: #Winkel der Peaks:
    theta_ka1 = np.radians(angle_4[peaks_4a[1]])
    theta_kb1 = np.radians(angle_4[peaks_4a[0]])
    theta_ka2 = np.radians(angle_4[46 + peaks_4b[1]])
    theta_kb2 = np.radians(angle_4[46 + peaks_4b[0]])

dtheta = np.radians(0.1)
```

```
[34]: #ergebnisse von teil 2
      beta_ka1 = np.radians(beta_ka1)
      beta_kb1 = np.radians(beta_kb1)
      beta_ka2 = np.radians(beta_ka2)
      beta_kb2 = np.radians(beta_kb2)
      dbeta = np.radians(dbeta)
[35]: d_{ka1} = d * np.sin(beta_{ka1})/np.sin(theta_{ka1})
      d_kb1 = d * np.sin(beta_kb1)/np.sin(theta_kb1)
      d_ka2 = d * np.sin(beta_ka2)/np.sin(theta_ka2)
      d_kb2 = d * np.sin(beta_kb2)/np.sin(theta_kb2)
      dd ka1 = d * np.sqrt((dbeta * np.cos(beta ka1)/np.sin(theta ka1))**2 + (dtheta_
      * np.sin(beta_ka1) * np.cos(theta_ka1) /(np.sin(theta_ka1)**2))**2)
      dd kb1 = d * np.sqrt((dbeta * np.cos(beta kb1)/np.sin(theta kb1))**2 + (dtheta
      * np.sin(beta_kb1) * np.cos(theta_kb1) /(np.sin(theta_kb1)**2))**2)
      dd ka2 = d * np.sqrt((dbeta * np.cos(beta ka2)/np.sin(theta ka2))**2 + (dtheta_1
      →* np.sin(beta_ka2) * np.cos(theta_ka2) /(np.sin(theta_ka2)**2))**2)
      dd kb2 = d * np.sqrt((dbeta * np.cos(beta kb2)/np.sin(theta kb2))**2 + (dtheta
       * np.sin(beta_kb2) * np.cos(theta_kb2) /(np.sin(theta_kb2)**2))**2)
[36]: #mittelwert:
      ds = np.array([d_ka1, d_ka2, d_kb1, d_kb2])
      dds = np.array([dd_ka1, dd_ka2, dd_kb1, dd_kb2])
      d_NaCl = np.mean(ds)
      dd_NaCl = np.sqrt((np.std(ds, ddof=1)/np.sqrt(4))**2 + (1/4 * np.sum(dds))**2)
      print("Netzebenenabstand NaCl d = ({} +/- {})m".format(d_NaCl, dd_NaCl))
     Netzebenenabstand NaCl d = (2.921979889517671e-10 +/- 4.9896772917326705e-12)m
[37]: #tabelle:
      thetas = np.degrees(np.array([theta_ka1, theta_kb1, theta_ka2, theta_kb2]))
      dthetas = np.full(4, 0.1)
      head2 = ['Ordn', 'a/b', 'theta', 'dtheta', 'beta', 'dbeta', 'd', 'dd']
      tab2 = zip(ordn, ab, thetas, dthetas, angles, dangles, ds, dds)
      print(tabulate(tab2, headers=head2, tablefmt="latex"))
     \begin{tabular}{rlrrrrrr}
     \hline
        Ordn & a/b &
                         theta &
                                   dtheta &
                                              beta &
                                                       dbeta &
                                                                         d &
     dd \\
     \hline
```

```
4.86525e-12 \\
           1 & b
                          6.2 &
                                     0.1 &
                                             9.3 &
                                                       0.1 & 2.88902e-10 &
                    &
     2.36649e-12 \\
                                     0.1 &
                                                       0.1 & 3.01363e-10 &
           2 & a
                         14.4 &
                                             20.9 &
                  &
     5.81023e-12 \\
          2 & b
                         12.8 &
                                     0.1 &
                                            18.5 &
                                                       0.1 & 2.88448e-10 &
                    &
     2.67843e-12 \\
     \hline
     \end{tabular}
[38]: #Avogadro:
     M_NaCl = 58.44e-3 \#kq
     rho_NaCl = 2.164e3 #kg/m^3
     NA = 0.5 * M_NaCl / (rho_NaCl * d_NaCl**3)
     dNA = NA * np.sqrt((3 * dd_NaCl/d_NaCl)**2)
     NA_{lit} = 6.0221e23
     print("Avogadrozahl N_A = ({} +/- {})1/mol".format(NA, dNA))
```

0.1 &

10.4 &

0.1 & 2.90079e-10 &

7.2 &

print("Literaturwert N\_A = ({})1/mol".format(NA\_lit))

&

1 & a

Avogadrozahl  $N_A = (5.4124174730035766e+23 +/- 2.772731255472224e+22)1/mol$  Literaturwert  $N_A = (6.0221e+23)1/mol$ 

```
[39]: sign_NA = np.abs(NA - NA_lit)/dNA
print("Sigma_N_A = {}".format(sign_NA))
```

 $Sigma_N_A = 2.1988518569665287$