

Astrochemistry Tutorial 2: Maxwell A. Fine 14880725

```
In [2]: # imports
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
from scipy.optimize import curve_fit

data_file = "Methanol_fluxes.rtf"

/home/afinemax/anaconda3/lib/python3.9/site-packages/scipy/_init__.py:146: UserWarning: A NumPy version >=1.16.5 and <1.23.0 is required for this version of SciPy (detected version 1.26.3)
    warnings.warn(f"A NumPy version >={np_minversion} and <{np_maxversion}"
```

```
In [2]: # read in the data
data = []
with open(data_file) as infile:
    counter = 0
    for line in infile:

        # Skip lines that start with '{', '}', '\'
        if line[0] == '{' or line[0]=='}' or line[0]== '\\':
            continue

        counter += 1
        if counter == 1 :
            unit = line
            continue

        if counter == 2 :
            unit = line
            continue

    # Split the line into columns using spaces as the delimiter
    columns = line.strip().split()

    # Append the columns to the data list
    data.append(columns)

    # Remove '\\' from each string and convert to float
cleaned_data = [[float(value.rstrip('\\')) for value in row] for row in data]

# Convert the list of lists to a NumPy array
print(unit)
data = np.array(cleaned_data, dtype=float)
```

(GHz) (K) (s-1) (K km s-1) \

1. frequencies (v in GHz)
2. energies of the upper level (Eu in K)
3. Einstein-A coefficients (Aul in s -1)
4. degeneracies (gu)

5. integrated intensities ($W = \int T_{mb} dv$ in K km s -1)
6. 1σ error associated to W

Q1:

Write a small program (per example in Python), to plot a diagram of $\ln(N_u/g_u)$ as a function of Eu

$$\frac{N_\mu}{g_\mu} = C \frac{(\nu^2 W)}{A_{ul} g_u}$$

c = 1942.75

```
In [4]: freq = data[:, 0]
eu = data[:, 1]
aul = data[:, 2]
gu = data[:, 3]
w = data[:, 4]
w_sigma = data[:, 5]

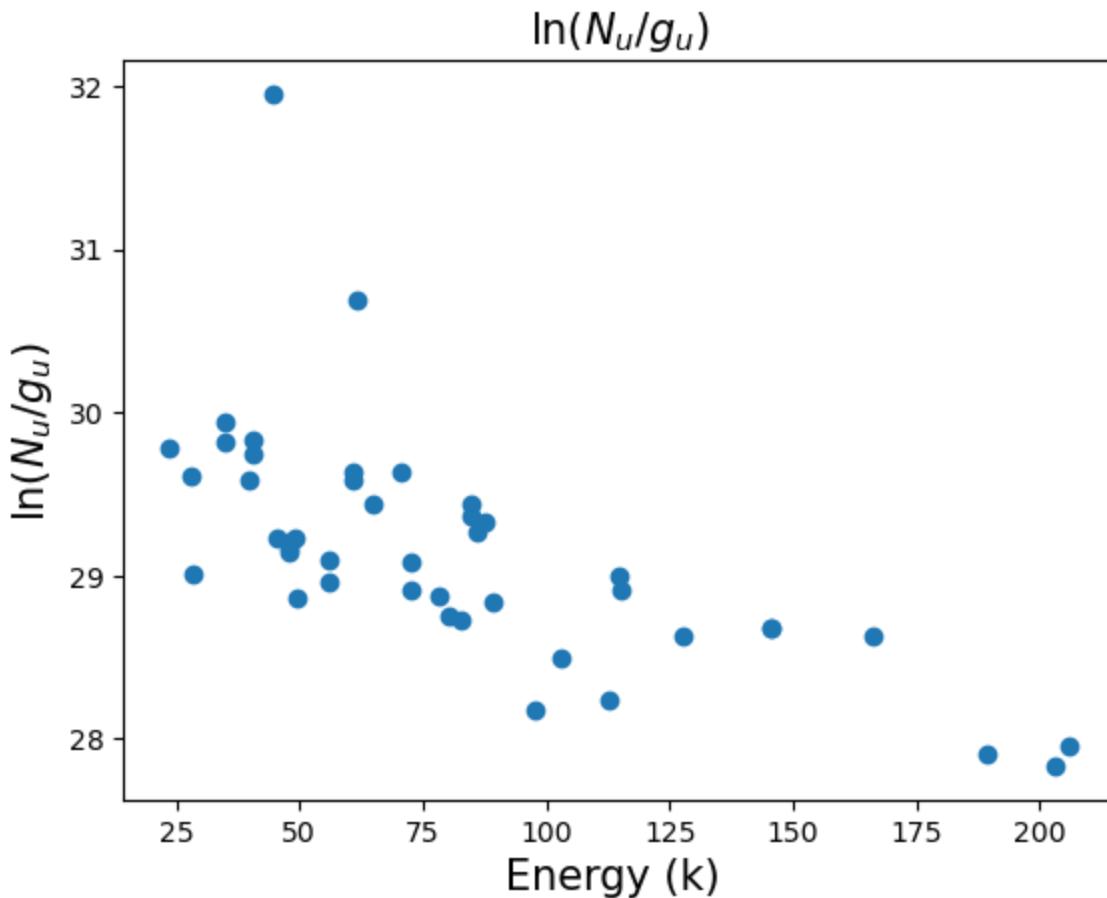
c = 19942.75

ratio = c * freq**2 * w / (aul * gu)

sigma_ratio = np.log(w_sigma/w * ratio)
sigma_ratio = w_sigma/w
ln_ratio = np.log(ratio)

plt.title('$\ln(N_u/g_u)$', size=15)
plt.scatter(eu, ln_ratio)
plt.xlabel('Energy (k)', size=15)
plt.ylabel('$\ln(N_u/g_u)$', size=15)
```

Out[4]: Text(0, 0.5, '\$\ln(N_u/g_u)\$')



Q2:

Q2: Fit a linear function to the data and obtain the values for a and b such that
 $\ln(N_u/g_u) = a + bE_u$

```
In [16]: # Define the linear function
def linear_function(eu, a, b):
    return a + b * eu

# Use curve_fit with w_sigma as the weights
popt, pcov = curve_fit(linear_function, eu, ln_ratio, sigma=sigma_ratio)

# Extract the values for a and b
a, b = popt

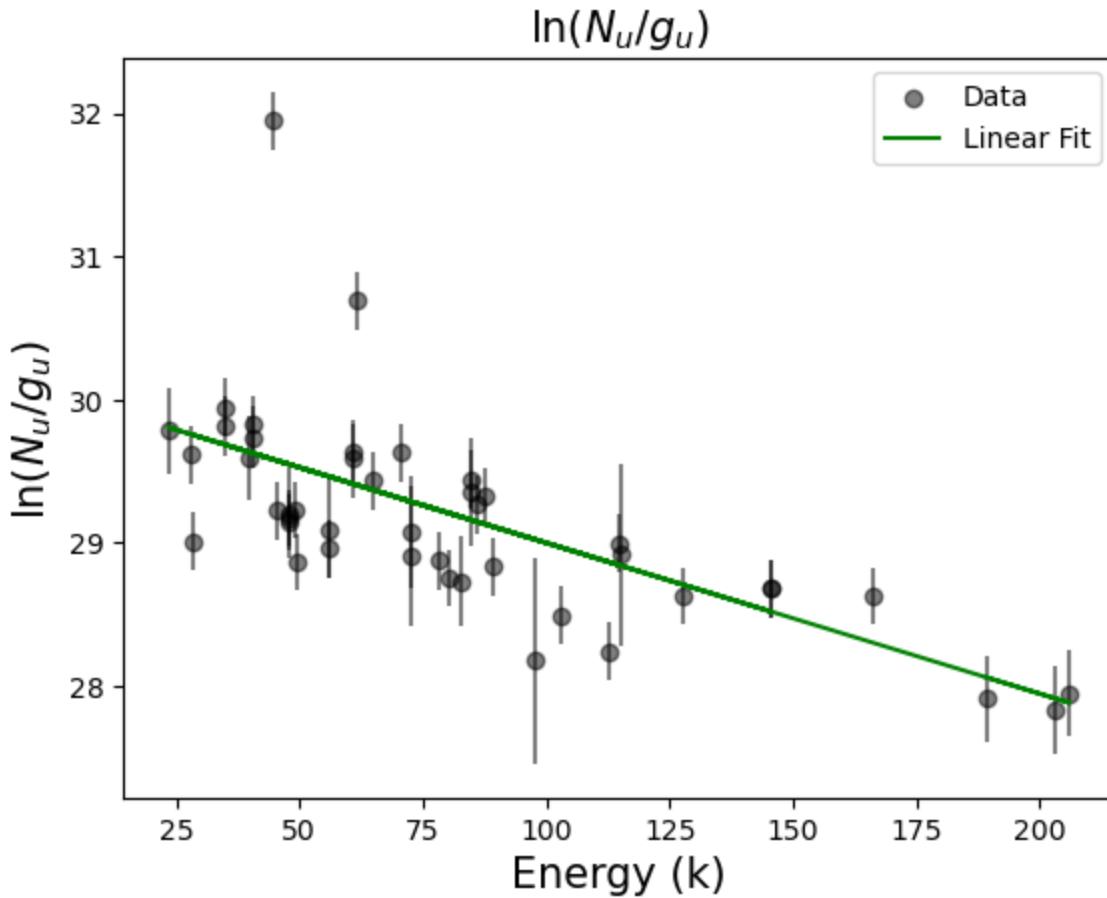
my_t = -1/b
my_a = a # saving a and b for later

# Print the values of a and b
print("a:", a)
print("b:", b)

# Plot the data and the fitted line
plt.title('$\ln(N_u/g_u)$', size=15)
plt.errorbar(eu, ln_ratio, yerr=sigma_ratio, fmt='None', color = 'k', alpha=0.5)
plt.scatter(eu, ln_ratio, color='k', label='Data', alpha=0.5)
plt.plot(eu, linear_function(eu, a, b), color='Green', label='Linear Fit') #label='Linear Fit'
plt.xlabel('Energy (k)', size=15)
```

```
plt.ylabel('$\ln(N_u/g_u)$', size=15)
plt.legend()
plt.show()
```

```
a: 30.054085921293527
b: -0.010534341369827421
```



Q3 & Q4:

Tutorial 2

Q3 Identify a, b

Q4

$$\ln\left(\frac{N_0}{g_0}\right) = a + bE_0 = \ln\left[\frac{N}{Q(T_{\text{rot}})} e^{-E_0/T_{\text{rot}}}\right]$$

$$\Rightarrow a + bE_0 = \ln\left(\frac{N}{Q(T_{\text{rot}})}\right) + -\frac{1}{T_{\text{rot}}} E_0$$

$b = -\frac{1}{T_{\text{rot}}}$

$a = \ln\left(\frac{N}{Q(T_{\text{rot}})}\right)$

$$\Rightarrow e^a \pm \frac{N}{Q(T_{\text{rot}})}$$

$N = Q(T_{\text{rot}}) e^a$

Q4 part 2:

We can fit a cubic function to the data table to determine our Q , based on our T , which we find using their relations to a , and b

In [24]:

```
# Given data
T_table = np.asarray([300.0, 225.0, 150.0, 75.0, 37.5, 18.75, 9.375]) #k
Q_table = np.asarray([6413.573, 4165.815, 2267.776, 801.678, 283.465, 100.207, 0.0])

# Define the cubic function
def cubic_function(T, a, b, c, d):
    return a * T**3 + b * T**2 + c * T + d

# Fit the cubic function to the data
params, covariance = curve_fit(cubic_function, T_table, Q_table)

# Extract the parameters
a, b, c, d = params

# Generate a smooth curve using the fitted parameters
T_smooth = np.linspace(min(T_table), max(T_table), 100)
Q_smooth = cubic_function(T_smooth, a, b, c, d)

# Plot the original data and the fitted curve
plt.scatter(T_table, Q_table, label='Data', color='k')
plt.plot(T_smooth, Q_smooth, label='Cubic Fit', color='red', linestyle=':')
plt.xlabel('T (k)', size=15)
plt.ylabel('Q', size=15)
plt.title('Determining Q from the Data Given')
plt.axvline(my_t, label='Our $T_{rot}$')
plt.legend()
plt.show()

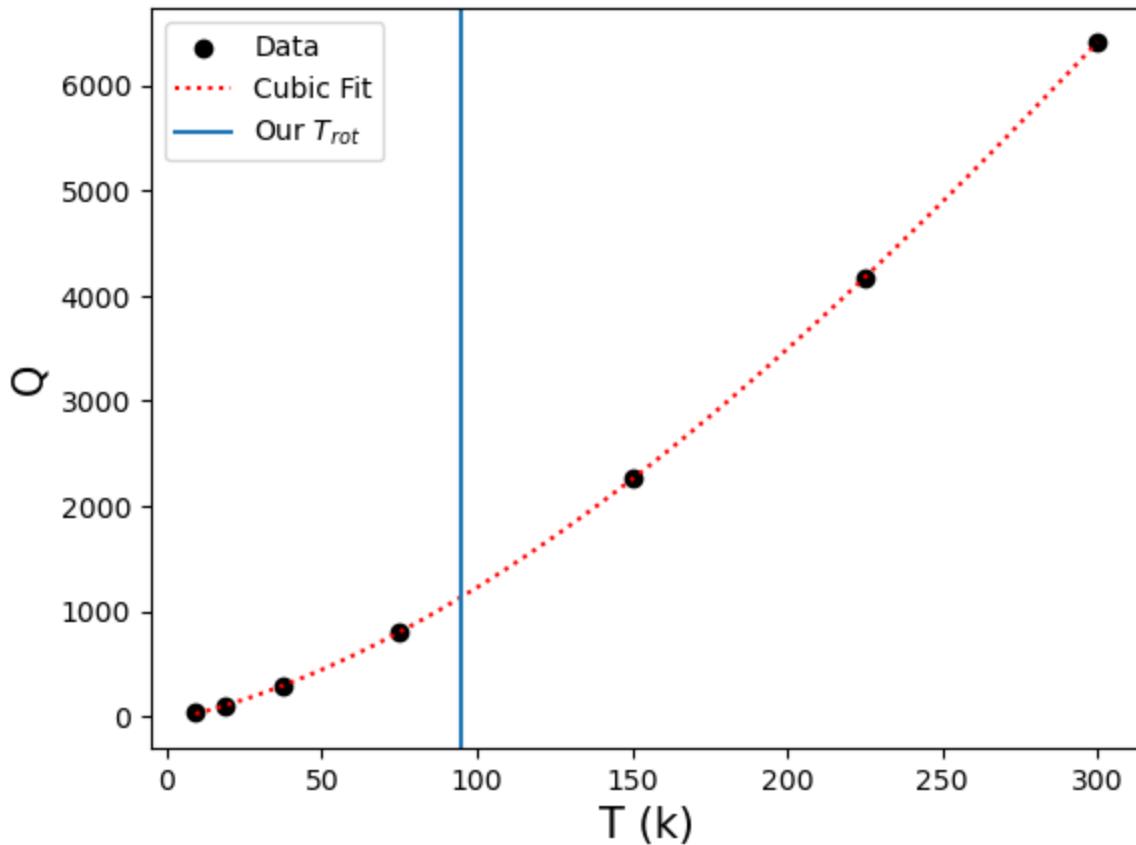
# Function to determine Q for a given T
def get_Q(T):
    return cubic_function(T, a, b, c, d)

# Example: Determine Q for a given T
resulting_Q = get_Q(my_t)
print('Answer for Q4')
print(f"For T = {my_t} k, Q = {resulting_Q}")

Q = resulting_Q
a = my_a
N = Q * np.exp(a)

print(f"N = {N}")
```

Determining Q from the Data Given



Answer for Q4

For $T = 94.92762431870788$ k, $Q = 1134.8932372414174$

$N = 1.2802025335819972e+16$

Ex 2

Tutorial 2 : Ex2

- Q1 1 - ~~Collision~~ Photo dissociation (Ar)
 2 Charge transfer
 3 Ion-molecule
 4 Radiative association
 5 Ion molecule
 6 Ion-molecule
 7 ~~Ion molecule~~ Ion-molecule
 8 Ion-molecule
 9 Ion-molecule
 10 ~~Ion molecule~~ dissociation / recombination
 11 photo dissociation

Possible Reaction types

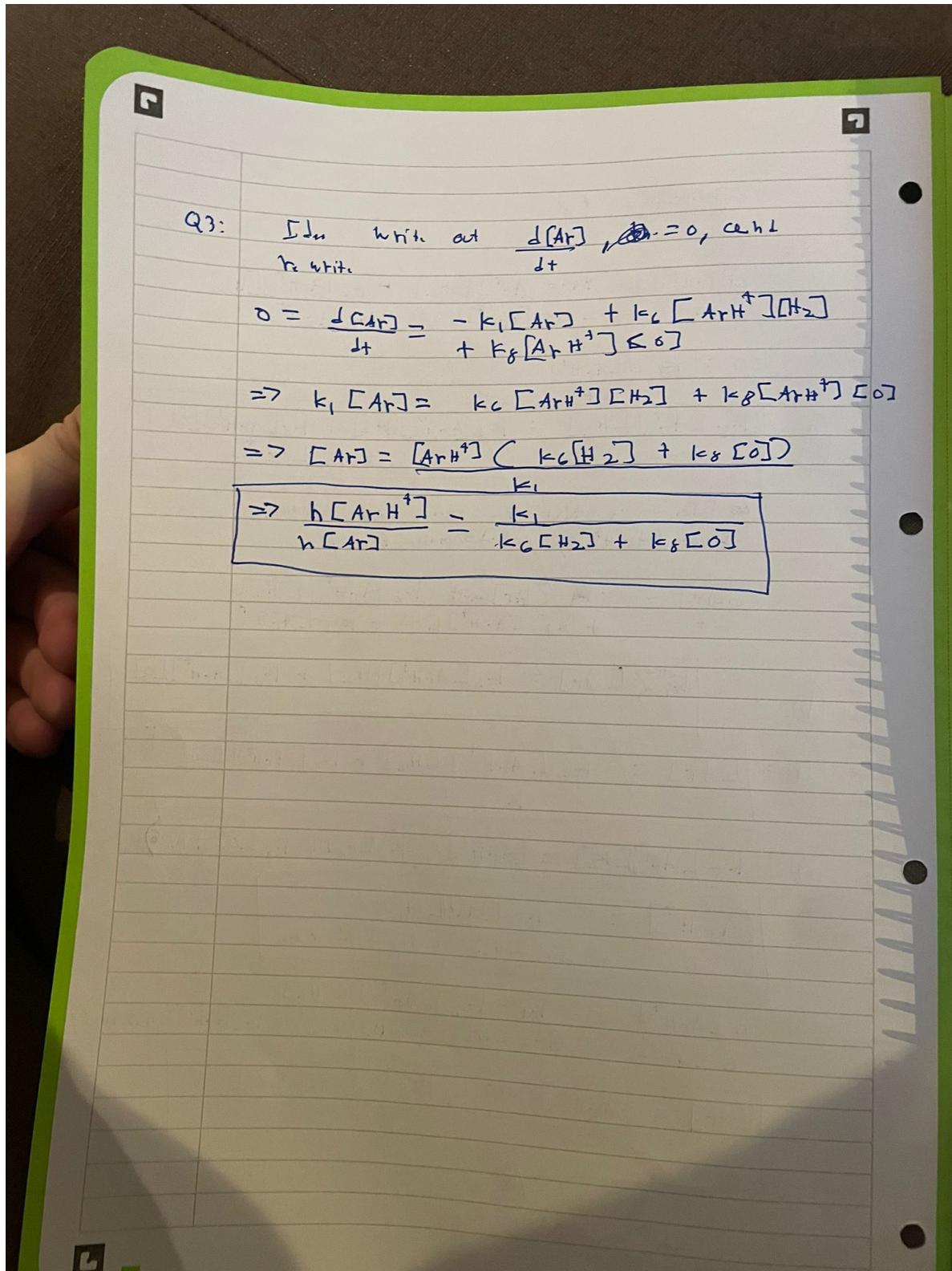
Q2

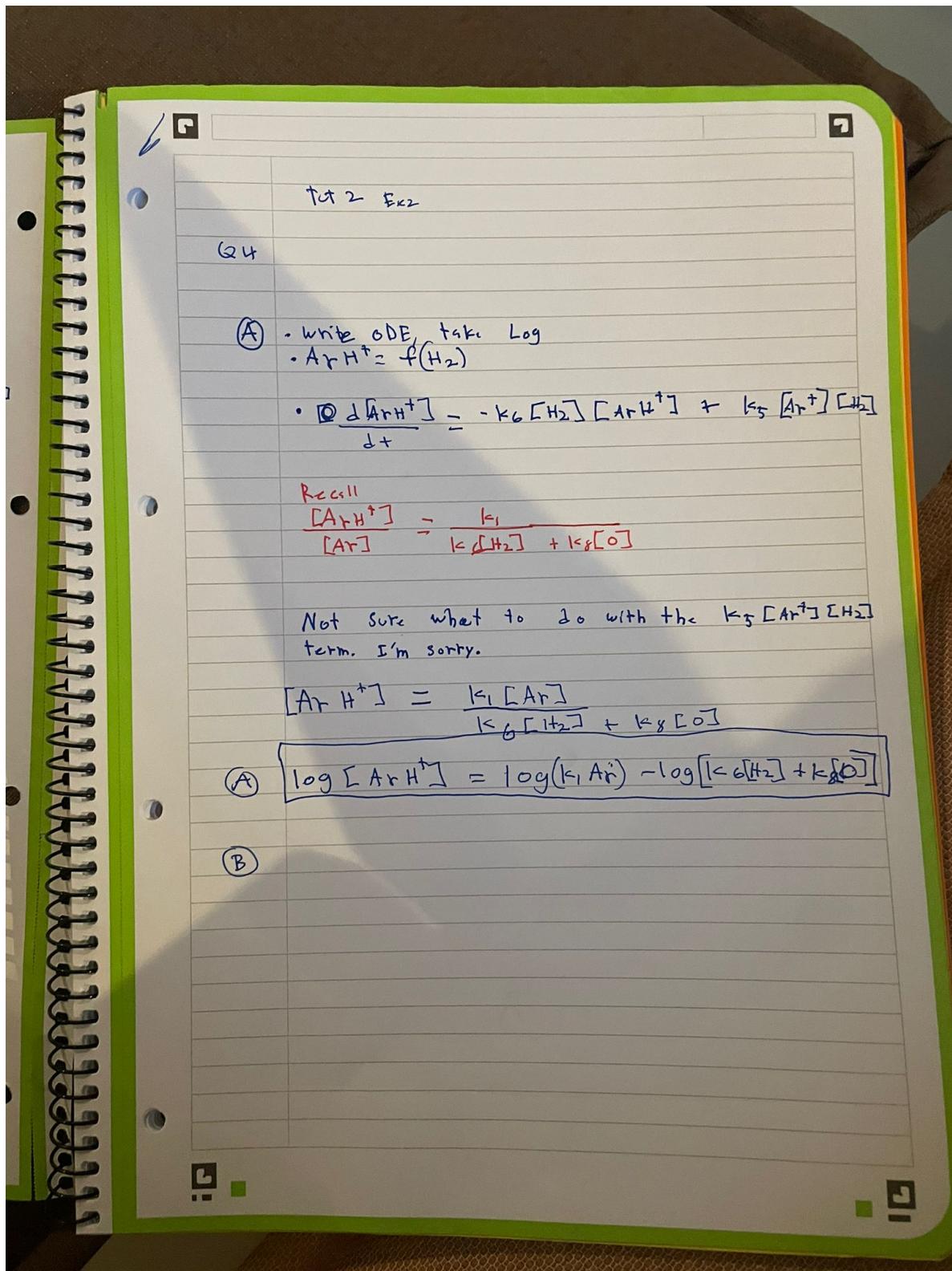


We write it as a sum of the reactions that create and destroy ArH^+ .

$$\frac{d[\text{ArH}^+]}{dt} = +k_3 [\text{Ar}] [\text{H}_3^+] + k_5 [\text{Ar}^+] [\text{H}_2] - k_6 [\text{ArH}^+] [\text{H}_2] - k_7 [\text{ArH}^+] [\text{CO}] - k_8 [\text{ArH}^+] [\text{O}] - k_9 [\text{ArH}^+] [\text{C}] - k_{10} [\text{ArH}^+] [\text{e}^-] - k_{11} [\text{ArH}^+] [\text{e}]$$

C
 Unimolecular, we
 do not include the
 photons in photo dissociation
 Reactions seen 3-66





Q4b

```
In [25]: #
def get_arh(h2, k1, k6, k8, o, ar):
    'handy function for calculating abundances'
    arh = k1 * ar / (k6*h2 + k8*o)
    return arh
```

```
In [29]: k1 = 11.4 * 2e-16
k6 = k8 = 1e-10

o = 2.9e-4
ar = 3.2e-6
h2 = np.asarray([-1, -2.5, -4])
h2 = 10**h2

my_arh = get_arh(h2, k1, k6, k8, o, ar)

print('abundance of ArH+')
print(my_arh)
```

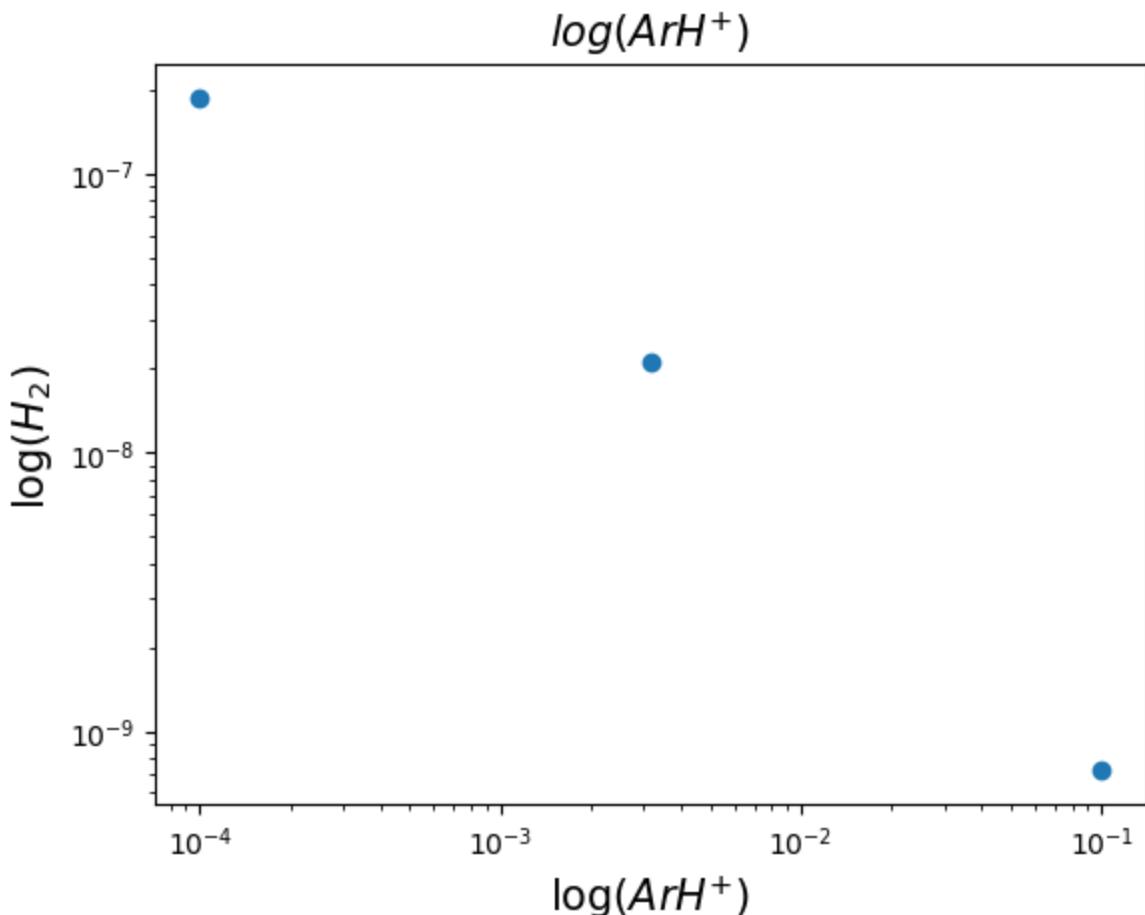
abundance of ArH+
[7.27490278e-10 2.11338737e-08 1.87076923e-07]

```
In [21]: plt.scatter(h2, my_arh)
plt.xscale('log')
plt.yscale('log')

plt.ylabel('$\log(H_2)$', size=15)
plt.xlabel('$\log(ArH^+)$', size=15)

plt.title('$\log(ArH^+)$', size=15)
```

Out[21]: Text(0.5, 1.0, '\$\log(ArH^+)\$')



Q4C

- From theory and from the plot, the $\log(ArH^+)$ is linear with a negative slope. The higher the abundance of ArH⁺ the lower the abundance of H₂
- we conclude that ArH⁺ is a good tracer of atomic gas because directly proportional to Ar, and inversely to H₂, (we derived the fraction ArH⁺/Ar already). Because of this relationship, we can observe ArH⁺ and infer the amounts of H₂, Ar, and O.

In []: