

09/14/20 04:14:04 /home/ana/Documents/uni/PHS3000/code/alpha\_particles.py

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1  # PHS3000
2  # alpha particles - Range and energy loss
3  # Ana Fabela, 09/09/2020
4  import os
5  from pathlib import Path
6  import monashpa.PHS3000 as spa
7  import numpy as np
8  import matplotlib.pyplot as plt
9  from scipy import special
10 import scipy.optimize
11
12 plt.rcParams['figure.dpi'] = 150
13
14 folder = Path('spectra')
15 os.makedirs(folder, exist_ok=True)
16
17 # Global prefixes and values SI units
18 x = np.linspace(0,1036,1024) # bins array
19
20 kilo = 1e3
21 cm = 1e-2 # [m]
22 g = 1e-3 # [kg]
23
24 eV = 1.602e-19 # [J]
25 MeV = eV * 1e6 # [J]
26 keV = eV * 1e3 # [J]
27
28 p_zero = 36.1 * kilo # [Pa]
29 u_p_zero = 0.5 * kilo
30
31 A_air = 14.5924
32 A_Au = 196.966570
33
34 rho_atm = 1.284 # [kg m**-3]
35 rho_gold = 19.30 * (g / cm**3) # [kg/ m3]
36
37 T = 294.15 # [K]
38 u_T = 0.5 # [K]
39
40 x_0 = 6.77 # [cm]
41 u_x_0 = 0.1 # [cm]
42
43 alpha_energy = 5.4857 # [MeV]
44
45
46 def read_files():
47     read_folder = Path('log2')
48     files = list(os.listdir(path=read_folder))
49     data_files = []
50     p_values = []
51
52     files.sort(key=lambda name: int(name.split('_')[0]) if name[0].isdigit() else -1)
53     for i, file in enumerate(files):
54         # print(i, file)
55         if i >= 1:
56             p_value = float(file.rstrip('mbar.mca').replace('_', '.'))
57             p_values.append(p_value)
58             header, data = spa.read_mca_file(read_folder/file)
59             data_files.append(data)
60     p_values = np.asarray(p_values)
61     u_p = np.sqrt((0.0025 * p_values)**2 + 0.01**2) # resolution and accuracy [kPa]
62     return p_values, u_p, data_files
63
64 def calibrate_axis(x, E_0, u_E_0):
65     # calibration factor for x-axis
66     _lbin = 5 * keV # [J]
67     u_lbin = (u_E_0 / E_0) * _lbin
68     # therefore x -> E
69     E = x * _lbin # [J]
70     u_E = (u_lbin / _lbin) * E
71     return E, u_E
72
73 def plot_data(x, y, i, average, ax=None):
74     xmax = np.argmax(y)

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75     ymax = y.max()
76     half_y_max = ymax / 2
77     L = np.argmin((y[:xmax] - half_y_max)**2)
78     R = np.argmin((y[xmax:] - half_y_max)**2) + xmax
79     peak_width = R - L
80
81     # if i >= 1:
82     #     text= "bin={:.0f}, count={:.0f}".format(xmax, ymax)
83     #     if not ax:
84     #         ax=plt.gca()
85     #         ax.annotate(text, xy=(xmax, ymax), xytext=(0.7, 1.02), textcoords='axes fraction')
86
87     # plt.bar(x, y, color='tomato',label="Detected alphas")
88     # plt.plot(x[:41], average, label="Extrapolation") # extrapolation cuve
89     # # plt.axhline(y=half_y_max, linestyle=':', alpha=0.3, label="Half max")
90     # # plt.fill_betweenx([0, ymax + 20], [L, L], [R, R], alpha=0.3, zorder=10)
91     # plt.xlim([0, 1024])
92     # plt.xlabel('Bins')
93     # plt.ylabel('Counts')
94     # plt.title(f'file[{i}].mca')
95     # plt.legend()
96
97     # spa.savefig(folder/f'file{i}.png')
98     # plt.show()
99     # plt.clf()
100     return xmax, ymax, peak_width
101
102 def energy_peaks(p_values, data_files):
103     max_counts = []
104     max_positions = []
105     peak_widths = []
106     total_events = []
107
108     for i, signal in enumerate(data_files):
109         # curve extrapolation for the threshold region
110         average_list = [np.mean(signal[42:92])] * 41
111         # sum of all events including extrapolation
112         total = np.around(np.sum(signal) + np.sum(average_list), decimals=0)
113         total_events.append(total)
114
115         # print(f"{np.sum(signal)} + {np.sum(average_list)} = {total}")
116
117         # barcharts to visualise our files
118         xmax, countmax, peak_width = plot_data(x, signal, i, average_list)
119
120         max_positions.append(xmax)
121         max_counts.append(countmax)
122         peak_widths.append(peak_width)
123     peak = max_positions[1]
124     return peak, signal, total_events, max_positions, max_counts, peak_widths
125
126
127 def plot_pressure_vs_energy(p_values, max_positions):
128     # pressure vs Energy peak
129     plt.plot(p_values, max_positions[1:], 'o', color='tomato', markersize=2.5, label="peak")
130     plt.xlabel('pressure / kPa')
131     plt.ylabel('Bin number')
132     plt.title(f' Position of peak vs pressure ')
133     plt.legend()
134     spa.savefig(f'peak_position_vs_pressure.png')
135     plt.show()
136
137 def Task_2(x_0, u_x_0):
138     # Calculating E_0
139     # from equation (5)
140     R_0 = x_0
141     u_R_0 = u_x_0
142     # print(f"\n{T = {:.2f}} K, R = {R_0} cm, p = {p_zero / kilo} kPa")
143     E_0 = (R_0 * p_zero / (64.31 * T))**(1 / 1.73)
144     u_E_0 = (E_0 / 1.73) * np.sqrt((u_R_0 / R_0)**2 + (u_p_zero / p_zero)**2 + (u_T / T)**2)
145     print(f"\nE_0 = {E_0:.2f} ± {u_E_0:.2f} MeV")
146     # # difference in energy
147     diff_range_E = alpha_energy - E_0
148     u_diff_range_E = u_E_0
149     print(f"\n{diff_range_E = {:.2f}} ± {u_diff_range_E:.2f}")
150     return R_0, u_R_0, E_0, u_E_0
151

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152 def Task_3(E_0, u_E_0, rho_atm, rho_gold, A_Au, A_air, alpha_energy):
153     # Calculating R_Au:
154     # equation (4)
155     R_atm = 0.186 * E_0**1.73
156     u_R_atm = R_atm * 1.73 * u_E_0 / E_0
157     # print(f"\nTheoretical Range in 1 atm: {R_atm = :.2f} ± {u_R_atm:.2f} cm")
158
159     # Calculating anticipated range for particles travelling through gold
160     # equation (13)
161     R_Au = R_atm * (rho_atm / rho_gold) * np.sqrt(A_Au / A_air)
162     u_R_Au = R_Au * (u_R_atm / R_atm)
163     # print(f"{R_Au = } ± {u_R_Au} cm")
164
165     # Calculating the thickness (delta_x) of the gold coating
166     # rearranged equation (14)
167     delta_x = np.sqrt(A_Au) * (E_0**1.73 - alpha_energy**(1/0.578)) / (4.464 * rho_gold * np.sqrt(A_air))
168     u_delta_x = 1.73 * u_E_0 / E_0
169     # print(f"\n{delta_x = } ± {u_delta_x} cm")
170     return R_atm, u_R_atm, R_Au, u_R_Au, delta_x, u_delta_x
171
172
173 def Task_4(E_0, u_E_0, R_atm, u_R_atm):
174     # theoretical plot of range vs energy of alpha particles in air at atm pressure
175     x, y = [E_0], [R_atm]
176     E_th = np.linspace(0, 6, 1024) # (0 - 6) MeV array
177     R_th = 0.186 * (E_th**1.73)
178
179     plt.plot(
180         E_th, R_th, marker="None",
181         linestyle="-", label=r"$R_{th}(E_0)$"
182     )
183     plt.errorbar(
184         x, y, xerr=u_E_0, yerr=u_R_0,
185         marker="None", linestyle="None", ecolor="tomato",
186         label=r"$R_{atm}$", color="tomato", barsabove=True
187     )
188
189     plt.xlabel(r'$E_0$ / MeV')
190     plt.ylabel('R / cm')
191     plt.title(r'Theoretical plot of R vs $E_0$ of alpha particles in air')
192     plt.legend()
193     spa.savefig(f'theoretical_R_vs_E.png')
194     plt.show()
195
196 def Task_5(peak_widths, p_values):
197     # plot of FWHM vs pressure
198     plt.plot(
199         p_values, peak_widths[1:], 'o', color='tomato', markersize=2.5, label=r"$FWHM(p)$"
200     )
201     plt.axvline(x=p_zero/kilo, linestyle='--', alpha=0.5, label="37 kPa")
202     plt.grid(linestyle=':')
203     plt.xlabel(r'$p$ / kPa')
204     plt.ylabel('FWHM')
205     plt.title(r'Width of energy spectra vs pressure')
206     plt.legend()
207     spa.savefig(f'FWHM_vs_pressure.png')
208     plt.show()
209
210 def Task_6(total_events):
211     # The effect of range straggling.
212     # How does the number of surviving particles vary with pressure.
213
214     plt.plot(
215         p_values, total_events[1:], 'o', markersize=2.5, color='tomato', label=r"detected $alphas$"
216     )
217     # plt.grid(linestyle=':')
218     plt.xlabel(r'$p$ / kPa')
219     plt.ylabel('detected particles')
220     plt.title(r'Surviving particles vs pressure')
221     plt.legend()
222     spa.savefig(f'alphas_vs_pressure.png')
223     plt.show()
224
225
226 def f(p_values, p_R, a):
227     # model for optimize.curve_fit()
228     return (1 / 2) * (1 - special.erf((p_values - p_R) / a))

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229
230 def Task_7_8(f, x, y):
231     p_0 = p_zero / kilo # [kPa]
232     u_p_0 = u_p_zero / kilo
233     # print(x)
234     # print(f"{p_0}")
235
236     # determining straggling parameter  $\alpha$ 
237     popt, pcov = scipy.optimize.curve_fit(f, x, y)
238     # To compute one standard deviation errors on parameter  $\alpha$ 
239     perr = np.sqrt(np.diag(pcov))
240
241     p_R, p $\alpha$  = popt
242     u_p_R, u_p $\alpha$  = perr
243
244     p = np.linspace(x[0], x[-1], 25)
245     optimal_fit = f(p, p_R, p $\alpha$ )
246
247     plt.plot(
248         x, y, marker='o', linestyle='None', markersize=2.5, color='tomato',
249         label=r"detected  $\alpha$ "
250     )
251     plt.plot(
252         p, optimal_fit, marker="None",
253         linestyle="-",
254         label="fit"
255     )
256
257     plt.grid(linestyle=':')
258     plt.xlabel(r'p / kPa')
259     plt.ylabel('Detected particles')
260     plt.title(r'Number alpha particles as a function of pressure')
261     plt.legend()
262     spa.savefig(f'alphas_vs_pressure.png')
263     # plt.show()
264     return p_R, p $\alpha$ , u_p_R, u_p $\alpha$ , p_0, u_p_0
265
266 def compare(p_R, p $\alpha$ , u_p_R, u_p $\alpha$ , p_0, u_p_0):
267     # conversion to distance units
268     p_atm = 101.325 # [kPa]
269     u_p_atm = np.sqrt((0.0025 * p_atm)**2 + 0.01**2) # [kPa]
270
271     x $\alpha$  = (p $\alpha$  / p_atm) * R_atm
272     u_x $\alpha$  = x $\alpha$  * np.sqrt((u_p $\alpha$  / p $\alpha$ )**2 + (u_p_atm / p_atm)**2 + (u_R_atm / R_atm)**2)
273     # comparison to theory
274     k = 0.015
275     diff_range = R_atm - x $\alpha$ 
276     how_many_sigmas = diff_range / u_x $\alpha$ 
277     print(f"\nEXPECTED RESULT  $\alpha \approx \{k * R_{atm} : .4f\} \pm \{k * u_{R_{atm}} : .4f\}$  cm")
278     print(f"Experimental  $\alpha = \{x_{\alpha} : .2f\} \pm \{u_{x_{\alpha}} : .2f\}$  cm")
279     # print(f"difference {diff_range:.3f}")
280     print(f"number of  $\sigma$  away from true result: {abs(how_many_sigmas):.3f}")
281
282     diff_p = p_0 - p_R
283     how_many_sigmas = diff_p / u_p_R
284     print(f"\nPrevious p_0: {p_0:.2f}  $\pm$  {u_p_0:.2f} kPa")
285     print(f"fit p_0 {p_R:.2f}  $\pm$  {u_p_R:.2f} kPa")
286     # print(f"difference {diff_p:.3f}")
287     print(f"number of  $\sigma$  away from true result: {abs(how_many_sigmas):.3f}")
288
289
290 ### * FUNCTION CALLS *###
291
292 p_values, u_p, data_files = read_files()
293
294 peak, signal, total_events, max_positions, max_counts, peak_widths = energy_peaks(p_values, data_files)
295
296 plot_pressure_vs_energy(p_values, max_positions)
297
298 R_0, u_R_0, E_0, u_E_0 = Task_2(x_0, u_x_0)
299
300 R_atm, u_R_atm, R_Au, u_R_Au, delta_x, u_delta_x = Task_3(E_0, u_E_0, rho_atm, rho_gold, A_Au, A_air,
301     alpha_energy)
302
303 E, u_E = calibrate_axis(x, E_0, u_E_0)
304
305 # Task_4(E_0, u_E_0, R_atm, u_R_atm)

```

```
305
306 # Task_5(peak_widths, p_values)
307
308 # Task_6(total_events)
309
310 y = total_events[1:] / total_events[1]
311 p_R, p $\alpha$ , u_p_R, u_p $\alpha$ , p_0, u_p_0 = Task_7_8(f, p_values, y)
312 # The straggling parameter can be expressed either as a pressure or a distance
313 # (at 1 atm pressure) it's just proportional to the range value in the units
314 # you've expressed it in.
315 print(f"\n{p $\alpha$  = {:.1f}  $\pm$  {u_p $\alpha$ :.1f} kPa")
316
317 compare(p_R, p $\alpha$ , u_p_R, u_p $\alpha$ , p_0, u_p_0)
318
319 # Which result do you think is more accurate and why?
320
321
```