▼ RP - Scintillation Detectors

Research Practicum: Scintillation Detectors Student names: Bouchra Bouras (5364213) and Zeno Hamers (5254981) Supervisors: Folkert Geurink, Dr.ir. Robin de Kruijff, Dr.ir. Antonia Denkova

Measurement days:

```
Day 1: 11/02/2022 (Well-Type: Sc-46, Cr-51, Co-57, Cs-137, Am-241; Coax-Type: Sc-46, Cr-51, Co-57, Cs-137, Tm-170)
```

```
Day 2: 14/02/2022 (Well-Type: Tm-170, Na-22; Coax-Type: Am-241, Na-22)
```

Research Objective: Analyzing the differences between NaI(TI) Well type and LaBr3(Ce) Coax type detector. Specifically energy calibration, absolute and relative energy resolution, and full energy peak resolution.

```
1 # Initialisation code for the notebook
2 import numpy as np
3 import matplotlib.pyplot as plt
4 from scipy.optimize import curve_fit
5 import math
```

Measurement Data:

```
1 #Data set (see corresponding spread sheet) #calculations to find R and FWHM can be foun
 3 #! NaI(TI) Well type detector data:
 4 #
 6 #Energy calibration data (all channel/energy data available)
 7 cal_channels_w = np.array([303,381,687,111,42,227,29,19,177,354,431,612,790])
 8 cal_energies_w = np.array([889.28,1120.552,2009.832,320.0835,129.267265,661.659,84.2547
 9 counts w = np.array([23532, 14979, 1691, 46325, 3525552, 82866, 1267485, 680426, 172608)
10
11 #Energy resolution data (only peaks with a Clean Gaussian form) #IMPORTANT! order of nu
12 name_w = np.array(['Sc-46 peak 1','Sc-46 peak 2', 'Cr-51', 'Co-57', 'Cs-137', 'Tm-170',
13 energies w = np.array([889.28,1120.552,320.0835,129.267265,661.659,84.25477,59.54092,51]
14
15 #Energy efficiency data (only peaks with a Clean Gaussian form)
16 energies_eff_w = np.array([889.28,1120.552,320.0835,129.267265,661.659,84.25477,59.5409
17 \text{ Pgamma}_{\text{w}} = \text{np.array}([99.98374, 99.97, 9.89, 85.49+10.71, 84.99, 2.48, 35.92, 99.94]) *
         = np.array([ 341.2988889,462.54, 581.9755556, 24377.50889, 1531.041111, 2331.005
18 R w
19
20
21 #
22 #1 12Dx2/Ca) Casy tuna datastan data.
```

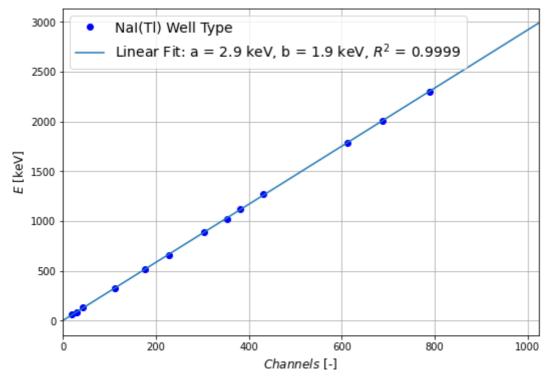
```
ZZ #! Labr3(ce) coax type detector data:
23 #
24
25 #Energy calibration data (all channel/energy data available)
26 cal_channels_c = np.array([293, 367, 652, 111, 49, 216, 37, 29, 166, 399, 557])
27 cal_energies_c = np.array([889.28, 1120.552, 2009.832, 320.0835, 129.267265, 661.659, 8
28 counts_c = np.array([24509, 16688, 565, 36570, 289316, 2708038, 52532, 710853, 706296,
29
30 #Energy resolution data (only peaks with a clean Gaussian form)
31 #IMPORTANT: order of nucleotides: Sc-46, Cr-51, Co-57, Cs-137, Tm-170, Am-241, Na-22
32 name_c = np.array(['Sc-46 peak 1','Sc-46 peak 2', 'Cr-51', 'Co-57', 'Cs-137', 'Tm-170',
33 energies_c = np.array([889.28, 1120.552, 320.0835, 129.267265, 661.659, 84.25477, 59.54
34
35
36 #Energy efficiency data (only peaks with a clean Gaussian form)
37 energies_eff_c = np.array([889.28, 1120.552, 320.0835, 129.267265, 661.659, 84.25477, 5
38 Pgamma_c = np.array([ 99.98374,99.97, 9.89, 85.49+10.71 , 84.99, 2.48, 35.92, 99.94]) *
39 R_c = np.array([238.2233333, 169.3811111, 199.1733333, 10255.80667, 434.6877778, 1402.6
40
41
42 #Background Radiation
43 counts_back_c = np.array([])
44
45 #error check:
46 def data_errorcheck(ar1,ar2):
47
    if len(ar1) == len(ar2):
48
      return True
49
    else:
50
       print("The arrays do not have the same length!")
52 data_errorcheck(name_w,energies_w)
53 data_errorcheck(name_c,energies_c)
54
    True
 1 #importing background radiation data
 2 background_w = np.loadtxt('/content/NaI(Tl)_BG.txt')
 3 background_c = np.loadtxt('/content/LaBr3(Ce)_BG.txt')
 4 background_w[0:2] = [0,0]
 5 background c[0:2] = [0,0]
```

Energy Calibration:

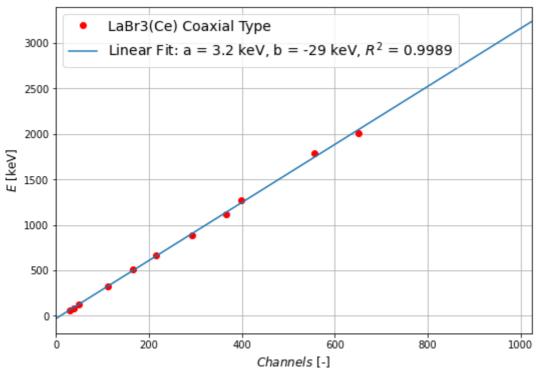
```
1
2 #curve fitting
3
4 def lin(x,a,b):
5    return a*x + b
6
7 x = np.linspace(0,1024,1000)
```

```
8
 9 var w, cov w = curve fit(lin, cal channels w, cal energies w)
10 var_c, cov_c = curve_fit(lin, cal_channels_c, cal_energies_c)
11
12 a_w = var_w[0]
13 b_w = var_w[1]
14
15 a_c = var_c[0]
16 b_c = var_c[1]
17
18 #comptuting R^2 value:
19
20 def R2(f, var, x, y):
   residuals = y - f(x,*var)
   ss_res = np.sum(residuals**2)
23
   ss\_tot = np.sum((y-np.mean(y))**2)
24
   r_squared = 1 - (ss_res/ss_tot)
25
   return r_squared
26
27 r_squared_w = R2(lin,var_w,cal_channels_w,cal_energies_w)
28 r_squared_c = R2(lin,var_c,cal_channels_c,cal_energies_c)
30 #plotting
31 #plt.subplots(figsize=(12,4))
32
33 #plt.subplot(121) #Well type
34 #plt.figure(figsize=(8,6))
35 plt.figure(figsize=(8.5,6))
36 plt.plot(cal_channels_w,cal_energies_w,'bo', label = "NaI(T1) Well Type")
37 plt.plot(x,lin(x,a_w,b_w), label = "Linear Fit: a = %.2g keV, b = %.2g keV, R^{2} = %
38 plt.xlabel("$Channels$ [-]", fontsize = 12)
39 plt.ylabel("$E$ [keV]", fontsize = 12)
40 plt.xlim(0,1024)
41 plt.grid()
42 plt.legend(fontsize = 14)
43 plt.savefig('energy_calibration_w.eps', format='eps')
44 plt.show()
45
46 #plt.subplot(122) #Coaxial type
47 plt.figure(figsize=(8.5,6))
48 plt.plot(cal_channels_c,cal_energies_c,'ro', label = "LaBr3(Ce) Coaxial Type")
49 plt.plot(x,lin(x,a_c,b_c), label = "Linear Fit: a = %.2g keV, b = %.2g keV, R^{2} = %
50 plt.xlabel("$Channels$ [-]", fontsize = 12)
51 plt.ylabel("$E$ [keV]", fontsize = 12)
52 plt.xlim(0,1024)
53 plt.grid()
54 plt.legend(fontsize = 14)
55 plt.savefig('energy calibration c.eps', format='eps')
56 plt.show()
57
58 print("a,b voor well type =", (a_w, b_w))
59 print("a,b voor coaxial type =", (a_c, b_c))
60
61
```

The PostScript backend does not support transparency; partially transparent artists ν The PostScript backend does not support transparency; partially transparent artists ν



The PostScript backend does not support transparency; partially transparent artists ν The PostScript backend does not support transparency; partially transparent artists ν



a,b voor well type = (2.915921421714314, 1.939967416979429)

a,b voor coaxial type = (3.1852987624416818, -28.663253251752604)

Discussion

The energy calibration of the well type and coaxial type show a linear fit respectively.

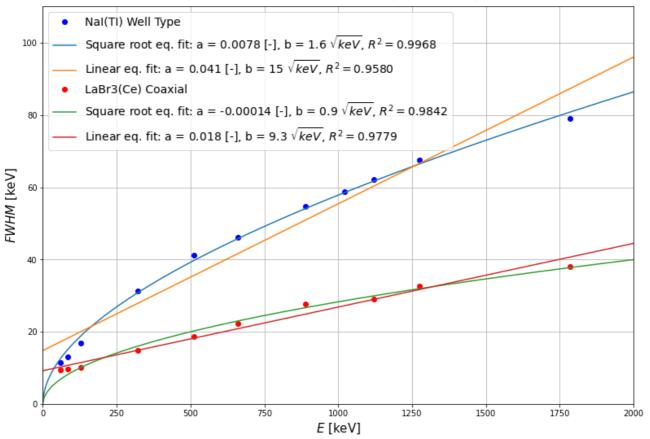
Comparing Energy Resolution

```
1 #plotting relative resolution against energy
 3 #argumentation why we take b = 0 for the calculation of the FWHM:
 4 \#In correspondence with the supervisor Folkert Geurink, it is decided to take b = 0 for
 5 #Since the calibrated offset b_c would create negative FWHM, the unphysical implication
 6
 7 #Well-Type:
 8 #order of peaks: 'Sc-46 peak 1', 'Sc-46 peak 2', 'Cr-51', 'Co-57', 'Cs-137', 'Tm-170', '
 9 FWHM w = a w * np.array([18.77996156, 21.27886882, 10.77752328, 5.788738976, 15.8370858
10
11 #Coax-Type:
12 #order of peaks: 'Sc-46 peak 1', 'Sc-46 peak 2', 'Cr-51', 'Co-57', 'Cs-137', 'Tm-170', '
13 FWHM_c = a_c * np.array([8.707246766, 9.126561225, 4.67983351, 3.234263792, 7.031472183)
14
15
16 #relative resolution:
17 \text{ W}_w = \text{FWHM}_w * (100/\text{energies}_w)
18 \text{ W}_c = \text{FWHM}_c * (100/\text{energies}_c)
19
20 \text{ index} = 4
21 print("Relative resolution W for %s = %.3g percent" %(name_w[index], W_w[index]))
22 print("FWHM for %s = %.3g keV" %(name_w[index], FWHM_w[index]))
23
24
25 #fitting empirical (square root) equation
26 def abs_res_squareroot(E,a,b):
                                                  #a and b have units of %
27
    return (a + b/np.sqrt(E))*E/100
28
29 E = np.linspace(0, 2000, 10000)
30 var_w,cov_w = curve_fit(abs_res_squareroot, energies_w, FWHM_w)
31 var_c,cov_c = curve_fit(abs_res_squareroot, energies_c, FWHM_c)
32
33 r_sq_w = R2(abs_res_squareroot,var_w,energies_w,FWHM_w)
34 r_sq_c = R2(abs_res_squareroot,var_c,energies_c,FWHM_c)
35
36 a_wf = var_w[0]
37 b_wf = var_w[1]
38 a_cf = var_c[0]
39 b_cf = var_c[1]
41 #fitting linear equation for the LaBr3(Ce) measurement data
42
43 #Well type:
44 var_wl, cov_wl = curve_fit(lin, energies_w, FWHM_w)
45 r sq wl = R2(lin, var wl, energies w, FWHM w)
46 \text{ a wl} = \text{var wl}[0]
47 b_wl = var_wl[1]
48
49 #Coaxial type:
50 var_cl, cov_cl = curve_fit(lin, energies_c, FWHM_c)
51 r_sq_cl = R2(lin,var_cl,energies_c,FWHM_c)
52 a_cl = var_cl[0]
53 b_cl = var_cl[1]
54
```

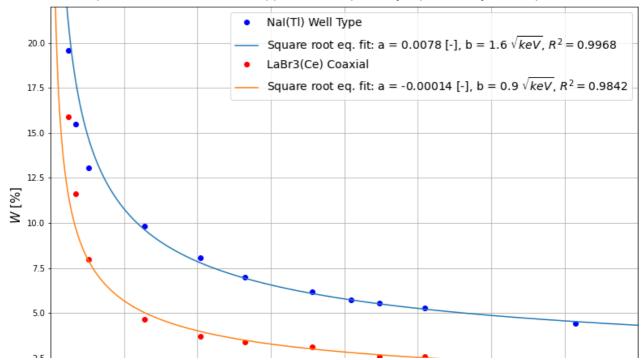
```
55
56
57 #comparing FWHM (absolute resolution):
58 plt.figure(figsize=(13,9))
59 plt.plot(energies_w,FWHM_w,'bo', label = "NaI(TI) Well Type")
60 plt.plot(E,abs_res_squareroot(E,a_wf,b_wf), label = "Square root eq. fit: a = %.2g [-],
61 plt.plot(E,lin(E,a_wl,b_wl), label = "Linear eq. fit: a = %.2g [-], b = %.2g $\sqrt{keV}
62 plt.plot(energies_c,FWHM_c,'ro', label = "LaBr3(Ce) Coaxial")
63 plt.plot(E,abs_res_squareroot(E,a_cf,b_cf), label = "Square root eq. fit: a = %.2g [-],
64 plt.plot(E,lin(E,a_cl,b_cl), label = "Linear eq. fit: a = %.2g [-], b = %.2g $\sqrt{keV}
66 plt.xlabel("$E$ [keV]", fontsize = 15)
67 plt.ylabel("$FWHM$ [keV]", fontsize = 15)
68 plt.xlim(0,2000)
69 plt.ylim(0,110)
70 plt.grid()
71 plt.legend(fontsize = 14)
72 plt.savefig('FWHM.eps', format='eps')
73 plt.show()
74
75
76 #comparing W (relative resolution):
77 #Well type
78 plt.figure(figsize=(13,9))
79
80 plt.plot(energies_w,W_w,'bo', label = "NaI(Tl) Well Type")
81 plt.plot(E,abs_res_squareroot(E,a_wf,b_wf)*100/E, label = "Square root eq. fit: a = %.2
82 plt.plot(energies_c,W_c,'ro', label = "LaBr3(Ce) Coaxial")
83 plt.plot(E,abs_res_squareroot(E,a_cf,b_cf)*100/E, label = "Square root eq. fit: a = %.2
84
85
86 plt.xlabel("$E$ [keV]", fontsize = 15)
87 plt.ylabel("$W$ [%]", fontsize = 15)
88 plt.xlim(0,2000)
89 plt.ylim(0,22)
90 plt.grid()
91 plt.legend(fontsize = 14)
92 plt.savefig('W.eps', format='eps')
93 plt.show()
```

Relative resolution W for Cs-137 = 6.98 percent FWHM for Cs-137 = 46.2 keV

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:27: RuntimeWarning: divi/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:27: RuntimeWarning: inva/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:27: RuntimeWarning: divi/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:27: RuntimeWarning: inva/local/lib/python3.7/dist-packages/ipykernel_launcher.py:27: RuntimeWarning: inva/local/lib/python3.7/dist-packages/ipykernel_launcher.py:27: RuntimeWarning: inva/local/lib/python3.7/dist-packages/ipykernel_launcher.py:27: RuntimeWarning: inva/local/lib/python3.7/dist-packages/ipykernel_launcher.py:27: RuntimeWarning: divi/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:27: RuntimeWarning: divi/usr/local/lib/python3



/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:27: RuntimeWarning: divi/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:27: RuntimeWarning: inva/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:27: RuntimeWarning: divi/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:27: RuntimeWarning: invaThe PostScript backend does not support transparency; partially transparent artists variety backend does not support transparency; partially transparent artists variety.



Discussion

For the relative energy resolution, a curve fit is needed to determine the relationship between energy and relative resolution. The deminishing decreasing trend points to exponential decay or inversely proportional relation. An inversely proportional relation would be expected because of the determining of the relative energy:

$$W = FWHM * 100/E.$$

The absolute energy resolution seems to be showing a linear relation, by a applying a curve fit, this relationship can be checked.

Comparing Full Energy Peak Efficiency

Computing activity:

```
1 from re import I
 2 #berekening huidige activiteit
 4 #Measurement day 1: 11/02/2022 -> A_d1
 5 #Measurement day 2: 14/02/2022 ->A_d2
 7 #functie voor activiteit
 8 def activity(A0,t_12,t):
 9
       return A0 *(1/2)**(t/t_12)
10
11 name_w = np.array(['Sc-46 peak 1', 'Sc-46 peak 2', 'Cr-51', 'Co-57', 'Cs-137', 'Tm-170',
12
13 #A0 radionucliden
14
15 #! NaI(TI) Well type detector:
16 # Day 1: Sc-46, Cr-51, Co-57, Cs-137, Am-24 # Day 2: Tm-170, Na-22
17 \text{ A0}_{w} = \text{np.array}([26.2e3, 67e3, 1.94e9, 15e3, 394e3, 47.4e3, 19.1e3]) \#Bq. Am-241 en Tm-
18 #volgorde: Sc, Cr, Co, Cs, Tm, Am, Na (correct order with measurement input data)
19
20 #! LaBr3(Ce) Coax type detector:
21 # Day 1: Sc-46, Cr-51, Co-57, Cs-137, Tm-170 # Day 2: Am-241, Na-22
22 \text{ A0}_{\text{c}} = \text{np.array}([26.2e3, 67e3, 1.94e9, 15e3, 394e3, 47.4e3, 19.1e3]) #Bq
23 #volgorde: Sc, Cr, Co, Cs, Tm, Am, Na (correct order with measurement input data)
24
25 #Times
26 import datetime as dt
27
28 #seconds in a day/year
29 \text{ dag} = 24*60*60 \text{ #s}
30 jaar = 365*dag #s
31
32 #Half-life in seconds
```

```
33 t_12_w = np.array([83.787*dag, 27.704*dag, 271.81*dag, 30.05*jaar, 127.8*dag, 432.6*jaa
34 t 12 c = t 12 w
35 #volgorde: Sc, Cr, Co, Cs, Tm, Am, Na (correct order with measurement input data)
36
37 #t_0 date of establishing A0 (chosen at 12 in the afternoon)
38 #t_d1 and t_d2 seconds between date of establishing A0 (t_0) and date of measurements (
39 d1 = dt.datetime(2022,2,11,14,30,00)
40 d2 = dt.datetime(2022, 2, 14, 9, 00, 00)
41
42 #times of activity determining
43 t_0_c = np.array([dt.datetime(2021,8,20,12,00,00),
                                                         #Sc-46
                     dt.datetime(2021,12,16,12,00,00),
                                                         #Cr-51
45
                     dt.datetime(2010,4,13,12,00,00),
                                                         #Co-57
46
                     dt.datetime(2014,11,17,12,00,00),
                                                         #Cs-137
47
                     dt.datetime(2022,2,1,12,00,00),
                                                         #Tm-170
48
                     dt.datetime(2020,2,4,12,00,00),
                                                        #Am-241
49
                     dt.datetime(2022,1,17,12,00,00)])
                                                         #Na-22
50 t_0_w = t_0_c
51
52 #make empty array for seconds between determined activity and measurement on day 1 and
53 t_d1_w = np.zeros(len(t_0_w))
54 t_d2_w = np.zeros(len(t_0_w))
55
56 #fill arrays for both days
57 for i in range(len(t 0 w)):
    t_d1_w[i] = (d1-t_0_w[i]).total_seconds()
59
    t_d2_w[i] = (d2-t_0_w[i]).total_seconds()
60
61 #remove radionuclides from days they were not measured.
62 # Day 1: Sc-46, Cr-51, Co-57, Cs-137, Am-241
63 # Day 2: Tm-170, Na-22
64 t_w = t_d1_w
65 t_w[4] = t_d2_w[4]
66 t_w[6] = t_d2_w[6]
67 #volgorde: Sc, Cr, Co, Cs, Tm, Am, Na (correct order with measurement input data)
68
69 #make empty array for seconds
70 t d1 c = np.zeros(len(t 0 c))
71 t_d2_c = np.zeros(len(t_0_c))
72
73 #fill arrays for both days
74 for i in range(len(t_0_c)):
   t_d1_c[i] = (d1-t_0_c[i]).total_seconds()
75
76
    t_d2_c[i] = (d2-t_0_c[i]).total_seconds()
77
78 #remove radionuclides from days they were not measured
79 t_d1_c = t_d1_c[:-2]
80 t d2 c = t d2 c[5:]
81 t_c = np.append(t_d1_c, t_d2_c)
82
83 #Activiteit
84 A_w = activity(A0_w, t_12_w, t_w)
85 A_c = activity(A0_c, t_12_c, t_c)
86
87 #Scandium has two peaks in its spectrum -> Activity needs to be first two elements in a
```

```
88 A_w = np.append(A_w[0], A_w)
89 A_c = np.append(A_c[0], A_c)
90
91 print("activity well type radionuclides in Bq", A_w, )
92 print("activity coaxial radionuclides in Bq", A_c)

activity well type radionuclides in Bq [ 6154.4168945 6154.4168945 16054.04095:
    12692.57473094 367425.63373431 47246.66592321 18715.48376095]
activity coaxial radionuclides in Bq [ 6154.4168945 6154.4168945 16054.04095344
    12692.57473094 372989.05398496 47246.09124462 18715.48376095]
```

Plotting Full Energy Peak Efficiency:

```
1 #e stands for "epsilon" and symbolizes the full-energy peak efficiency
2 #activity A
3 #P is the probability of emission of the particular photon being measured
4 e_w = R_w/(A_w * Pgamma_w)
5 e_c = R_c/(A_c * Pgamma_c)
6
7
8 plt.figure(figsize=(8.5,6))
9
10 plt.plot(energies_eff_w, e_w, "bo", label="NaI(TI) Well Type")
11 plt.plot(energies_eff_c, e_c, "ro", label="LaBr3(Ce) Coaxial")
12 plt.ylabel("Full Energy Peak Efficiency [-]", fontsize = 14)
13 plt.xlabel("$E$ [keV]", fontsize = 14)
14 plt.grid()
15 plt.legend(fontsize = 14)
16 plt.savefig('FEPE.eps', format='eps')
17 plt.show()
```

Discussion

The difference between the two detectors becomes apparent in these graphs. For both detectors, the maximum efficiency is reached for co-57, with a combined energy of both peaks of 129,267265 keV. While the energy is the same, the maximum peak efficiency is different for both scintillation detectors. It comes down to 0.8 for the well type and 0.34 for the coaxial type respectively.

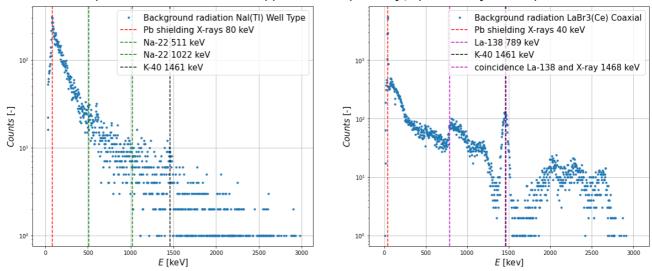
y |

Background Radiation

```
1 #compute the corresponding energies with the measured channels
 3 backchannels = np.linspace(0,1023,1024)
 5 E_back_w = a_w * backchannels + b_w
 7
 8 \# Argumentation why a_c = 3:
 9 #The background radiation was measured on two occasions,
10 #the first of which only a print-screen was made and
11 #this was not conclusive on the origin of the background radiation.
12 #For this reason another background radiation measurement was performed approximately 1
13 #Because of the change in environmentally factors between the two background radiation
14 #the choice was made to adjust the energy calibration constant $a$ to 3 keV.\\
15
16 a_c = 3
17 E_back_c = a_c * backchannels + b_c
19 #plotting the background radiation counts vs energies
20 plt.subplots(figsize=(20,8))
21
22 plt.subplot(121) #Well type
23 plt.plot(E_back_w, background_w,'.', label = "Background radiation NaI(Tl) Well Type")
24 plt.xlabel("$E$ [keV]", fontsize = 15)
25 plt.ylabel("$Counts$ [-]", fontsize = 15)
26 plt.yscale("log")
27 plt.axvline(x = 80, c = "r", linestyle = '--', label = "Pb shielding X-rays 80 keV")
28 plt.axvline(x = 511, c = "g", linestyle = '--', label = "Na-22 511 keV")
29 plt.axvline(x = 1022, c = "g", linestyle = '--', label = "Na-22 1022 keV")
30 #plt.axvline(x = 662, c = "r", linestyle = '--', label = "Ce-137 662 keV")
31 plt.axvline(x = 1461, c = "k", linestyle = '--', label = "K-40 1461 keV")
32 #plt.xlim(0,1024)
33 plt.grid()
34 plt.legend(fontsize = 15)
35
36
37 plt.subplot(122) #Coaxial type
38 plt.plot(E_back_c, background_c,'.', label = "Background radiation LaBr3(Ce) Coaxial")
39 plt.xlabel("$E$ [kev]", fontsize = 15)
```

```
40 plt.ylabel("$Counts$ [-]", fontsize = 15)
41 plt.axvline(x = 40, c = "r", linestyle = '--', label = "Pb shielding X-rays 40 keV")
42 plt.axvline(x = 789, c = "m", linestyle = '--', label = "La-138 789 keV")
43 plt.axvline(x = 1461, c = "k", linestyle = '--', label = "K-40 1461 keV")
44 plt.axvline(x = 1468, c = "m", linestyle = '--', label = "coincidence La-138 and X-ray
45
46 plt.yscale("log")
47 #plt.xlim(0,1024)
48 plt.grid()
49 plt.legend(fontsize = 15)
50 plt.savefig('background_radiation.eps', format='eps')
51 plt.show()
```

The PostScript backend does not support transparency; partially transparent artists v The PostScript backend does not support transparency; partially transparent artists v The PostScript backend does not support transparency; partially transparent artists v The PostScript backend does not support transparency; partially transparent artists v



Materials:

Nal(TI):

- Na-22 peak at: 511 kev
- K-40 (0.0117% of natural K) decay: electron capture (1461 keV) to Ar-40 (in general, Nal(Tl) detectors have very little internal contamination compared to LaBr3(Ce))

LaBr3(Ce):

- x- ray peak at 30-40 keV
- Ac-227 (with daughter La-138)
- Ac-227 dominates spectrum 1.6-3 MeV (purification is more successful for this radioisotope)
- La-138 dominates below 1.6 MeV
- La-138 (789-keV)
- 789 kev gamma ray in coincidence with a beta particle?
- also 1435 keV peak that becomes a 1468 keV peak under x-ray interactions. (http://www.nucleide.org/Laraweb/Result_Lara2.php)
- K-40 (1461 keV)

Investigating detector geometry differences

Discussion

The geometric differences have the greatest effect when annihilation takes place. As the well type encapsulates the radionuclide source. And for the coaxial type, the source lays on top. Annihilation happens when an electron and positron react with eachother and emit two photons in opposing directions. In the welltype there is a chance that both of these photons will be detected, while the coaxial type can detect one at the most.

×