Z Boson Lab Script

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```
Title: PHYS20161 - Assignment 2: Z Boson
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5
   Main Goal:
6
   This script is designed to analyze two particle collision data.
       First, it need
   to combine, validate the data sets. Second, it performs best fit
   finding the minimum of the chi_square value. Meanwhile, this code
      is able to
   get the parameters value for best fit and life time for ZO, and
       calculate the
   uncertainties. Finally, this code also have several additional
       features listed
   below.
12
13
   Additional Features:
14
   1. Further validation on the cross_section datas.
16
17
   2. Chi-Square Contour Plot: Provides a visualization of the
18
   chi_square minimization process.
19
   3. Extra 3D plot of the Contour Plot, optional output decided by
21
       the user.
22
   4. Validation whether the distribution is gaussian and the jusdge
23
       the quality
   of the fit: Estimate the geometric centre of the chi_square cntour
24
   define a funtction to assess the quality of the fit.
25
   (The user will see result clearly in output)
26
27
   5. Plots of the Gaussian distribution for both the mass (mass_z0)
       and the
   partial width ( ZO ).
29
   (To decide whether it is gaussian is checked by No.4 addtional
30
       features)
31
   6. CSV Data Export Function: A function to save filtered data for
       further use
```

```
or record-keeping.
33
   7. Interactive Menu: Allows users to customize the outputs based on
35
   preferences for further output options, enhancing user experience.
36
37
   Last updated: 13/12/2023
   @author: w23058yf
39
40
41
   import matplotlib.pyplot as plt
42
   import numpy as np
43
  import scipy.constants as pc
44
  from scipy.optimize import fmin
46
   # Constants
  | HBAR_GEV_S = pc.hbar / (1.602e-19 * 1e9) # Reduced Planck constant
48
        in GeV*s
   PARTIAL_WIDTH_EE = 83.91 * 1e-3 # Convert MeV to GeV
49
50
51
   # File names
  FILE_NAME_1 = 'z_boson_data_1.csv'
52
53
  FILE_NAME_2 = 'z_boson_data_2.csv'
54
55
   def validate_data(data):
56
57
       Validates the data by removing NaN values, zero or negative
58
           values,
59
60
       Args:
       data (array): The data array.
61
62
       Returns:
63
       array: The validated and cleaned data array.
64
65
66
       # Remove the NaN values
67
       data = data[~np.isnan(data).any(axis=1)]
68
69
       # Remove zero or negative values
70
71
       data = data[np.all(data > 0, axis=1)]
72
       return data
73
74
75
   def read_and_combine_data(file1, file2):
76
77
       Reads data from two chi_squared_contour_plotV files, validates,
78
            and
       combines them into a single array.
79
80
81
       Args:
       file1 (string): The first data file.
82
       file2 (string): The second data file.
83
84
85
       Returns:
```

```
array: A combined and validated array with data from both files
86
        or None if an error occurs.
87
88
89
        try:
            data1 = np.genfromtxt(file1, delimiter=',', skip_header=1)
90
            data2 = np.genfromtxt(file2, delimiter=',', skip_header=1)
91
92
            data1 = validate_data(data1)
93
            data2 = validate_data(data2)
94
95
            # Check column consistency between the two data sets
96
            if data1.shape[1] != data2.shape[1]:
97
                 raise ValueError(
98
                     "Mismatch in the number of columns between datasets
99
                         .")
100
            # Combine two sets of data sets
            combined_data = np.vstack((data1, data2))
            return combined_data
        except IOError as error:
103
            print(f"File not found: {error.filename}")
104
            return None
        except ValueError as error:
106
            print(f"Error reading file: {error}")
107
            return None
108
109
    def filter_data(data, num_std=3):
112
        Filter data by removing outliars based on a specified number of
113
114
        standard deviations.
115
116
        data (array): The data array.
117
        num_std (int): The number of standard deviations.
118
119
        Returns:
120
121
        array: The filtered data array.
122
123
        # Calculate the mean of the cross-section values
        mean_cross_section = np.mean(data[:, 1])
125
        # Calculate the standard deviation of the cross-section values
126
        std_dev_cross_section = np.std(data[:, 1])
127
128
        # Filter out data points where the cross-section deviates from
129
            the mean by
        # more than 'num_std' standard deviations.
130
        return data[abs(data[:, 1] - mean_cross_section) <= num_std *</pre>
131
                     std_dev_cross_section]
132
133
134
135
    def min_chi_square(center_of_mass_energy, cross_section,
        uncertainty,
136
                        initial_guesses):
137
138
        Find the parameters that minimize the chi-square value.
```

```
139
140
        center_of_mass_energy (array): Array of center-of-mass energies
141
        cross_section (array): Array of cross sections.
142
        uncertainty (array): Array of uncertainties in the cross
143
            section
        measurements.
144
        initial_guesses (list): List of initial guesses for the mass
145
            and partial
        width of the ZO boson.
146
147
        Returns:
148
149
        result(tuple)
150
        result = fmin(lambda parameters: chi_square(parameters,
                                                       center_of_mass_energy
153
                                                       cross_section,
                                                           uncertainty),
                       initial_guesses, full_output=True, disp=False)
        return result
156
157
    def filter_data_2(center_of_mass_energy, cross_section,
158
                       mass_z0_fit , partial_width_z0_fit ,
159
                           filtered_data_1):
160
        (This function is for No.1 additional feature)
161
163
        Filter data based on predicted cross sections compared to
            actual
        cross sections.
164
165
166
167
        Args:
        center_of_mass_energy (array): Array of center-of-mass energies
168
        cross_section (array): Array of measured cross sections.
169
170
        uncertainty (array): Array of uncertainties
        mass_z0_fit (float): Fitted Z0 boson mass.
171
172
        partial_width_z0_fit (float): Fitted ZO boson partial width.
173
        filtered_data_1 (array): Previously filtered data array.
174
        Returns:
        array: Further filtered data array.
176
177
178
        # Perdict the value of cross section by the formean_valuela.
        perdict_cross_section = calculate_cross_section(
179
             center_of_mass_energy, mass_z0_fit, partial_width_z0_fit)
        # Get the standard ddeviation value of the cross section.
181
        std_dev_cross_section = np.std(perdict_cross_section)
182
183
        return filtered_data_1[np.abs(perdict_cross_section -
            cross_section) <=</pre>
184
                                 0.5 * std_dev_cross_section]
185
186
```

```
def calculate_reduced_chi_square(chi_sq, num_data_points,
187
        num_parameters):
188
        Calculate the reduced chi-square value.
189
190
        Args:
191
192
        chi_sq (float): The chi-square value from the fit.
        num_data_points (int): The number of data points used in the
193
            fit.
        num_parameters (int): The number of parameters.
194
195
196
        Returns:
        float: The reduced chi-square value.
197
198
        degrees_of_freedom = num_data_points - num_parameters
199
        return chi_sq / degrees_of_freedom
200
201
202
    def calculate_cross_section(center_of_mass_energy, mass_z0,
203
        partial_width_z0):
204
        Calculate the cross section for a given center-of-mass energy,
205
        ZO boson mass, and partial width.
206
207
208
        Args:
        center_of_mass_energy (array): Array of center-of-mass energies
209
        mass_z0 (float): Z0 boson mass.
210
        partial_width_z0 (float): Z0 boson partial width.
211
212
213
        Returns:
        array: Calculated cross section values.
214
215
        # Define the given formula to calculate the cross section
216
        numerator = 12 * np.pi * center_of_mass_energy**2 * (
217
            PARTIAL_WIDTH_EE **2)
        denominator = (center_of_mass_energy**2 - mass_z0**2)**2 + \
218
219
             (mass_z0**2) * (partial_width_z0**2)
        return numerator / (denominator * mass_z0**2) * 389400
220
221
222
    def chi_square(parameters, center_of_mass_energy,
223
        cross_section_data,
                   uncertainty):
224
225
        Calculate the chi-square value for the fit.
226
227
        Args:
228
        parameters (list): List containing the mass and partial width
229
            of the ZO
230
        center_of_mass_energy (array): Array of center-of-mass
231
232
        energies.
        cross_section_data (array): Array of measured cross
233
234
        sections.
        uncertainty (array): Array of uncertainties
235
236
```

```
Returns:
237
        float: The calculated chi-square value.
238
239
        mass_z0, partial_width_z0 = parameters
240
        \# Get the perdiction value of cross section
241
        prediction = calculate_cross_section(center_of_mass_energy,
242
            mass_z0,
                                               partial_width_z0)
243
        # Get the chi_square value
244
        chi_sq = np.sum(((cross_section_data - prediction) /
245
            uncertainty) ** 2)
246
        return chi_sq
247
248
    def plot_data_and_fit(center_of_mass_energy, cross_section_data,
249
        uncertainty,
250
                           mass_z0, partial_width_z0):
251
252
        Plots the data points with error bars and the best fit curve.
253
        center_of_mass_energy (array): Array of center-of-mass energies
255
256
        cross_section_data (array): Array of measured cross sections.
        uncertainty (array): Array of uncertainties
257
        mass_zo (float): Z boson mass parameter
258
        partial_width_z0(float): Z boson partial width parameter
260
        Returns:
261
        None
262
263
        plt.figure(figsize=(10, 6))
264
        plt.errorbar(center_of_mass_energy, cross_section_data, yerr=
265
            uncertainty.
                      fmt='o', label='Data', color='royalblue', ecolor='
266
                          lightgray',
                      elinewidth=3, capsize=0)
267
268
        # Create a smooth line for the fit
269
        center_of_mass_energy_fit = np.linspace(
270
271
            min(center_of_mass_energy), max(center_of_mass_energy),
        cross_section_fit = calculate_cross_section(
272
            center_of_mass_energy_fit , mass_z0 , partial_width_z0)
273
274
        # Plot the fit
275
        plt.plot(center_of_mass_energy_fit, cross_section_fit,
276
                  label='Fit', linewidth=2, color='darkorange')
277
        plt.xlabel('Centre-of-mass energy (GeV)', fontsize=14)
278
        plt.ylabel('Cross Section (nb)', fontsize=14)
        plt.title('Fitted Curve with Scattered Data', fontsize=16)
280
        plt.legend()
281
        plt.grid(True)
282
        plt.savefig('z_boson_fit.png')
283
284
        plt.show()
285
286
```

```
def plot_chi_square_contours(center_of_mass_energy, cross_section,
287
        uncertainty,
                                   mass_z0_range, partial_width_z0_range,
288
                                   mass_z0_fit, partial_width_z0_fit):
289
290
        (This function is for No.2 additional feature)
291
292
        Provide visualization of chi_square for 2 parameters with fit
293
            values,
        and contour line which is helpful in future analysis.
294
295
        Args:
296
        center_of_mass_energy (array): Array of center-of-mass energies
297
         cross_section (array): Array of measured cross sections.
298
        uncertainty (array): Array of uncertainties in the cross
299
            section
        measurements.
300
        mass_z0_range (tuple): Range for the ZO boson mass range.
301
        partial_width_z0_range (tuple): Range for the ZO boson width
302
            range.
        mass_z0_fit (float): Fitted value of Z0 boson mass.
303
        partial_width_z0_fit (float): Fitted value of Z0 boson width.
304
305
        Returns:
306
        None
307
308
        mass_z0_values = np.linspace(*mass_z0_range, 100)
309
        partial_width_z0_values = np.linspace(*partial_width_z0_range,
310
        mass_z0_grid, partial_width_z0_grid = np.meshgrid(
            mass_z0_values, partial_width_z0_values)
312
         chi_sq_grid = np.zeros_like(mass_z0_grid)
313
314
        # Calculate chi-square values
315
316
        for i in range(mass_z0_grid.shape[0]):
            for j in range(mass_z0_grid.shape[1]):
317
318
                 chi_sq_grid[i, j] = chi_square([mass_z0_grid[i, j],
                                                   partial_width_z0_grid[i
319
                                                       , j]],
320
                                                  center_of_mass_energy,
                                                  cross_section,
321
                                                      uncertainty)
        # Find the minimum chi-square value
322
        best_chi_square = np.min(chi_sq_grid)
323
        level_1 = best_chi_square + 1
324
         # Plot Contours
325
        plt.contourf(mass_z0_grid, partial_width_z0_grid,
326
                      chi_sq_grid, levels=50, cmap='viridis')
327
        plt.colorbar(label='Chi-squared')
        # Add contour line
329
        {\tt plt.contour(mass\_z0\_grid\,,\,partial\_width\_z0\_grid\,,}
330
331
                     chi_sq_grid, levels=[level_1],
                     colors='red', linestyles='dashed')
332
        # Label for minimum chi-square + 1
333
        plt.plot([], [], 'r--', label='Min chi-square + 1')
334
335
        # Plotting the best fit values as a red dot
```

```
plt.plot(mass_z0_fit, partial_width_z0_fit, 'ro', label='
336
             Minimum Fit')
        plt.axhline(y=partial_width_z0_fit, color='r', linestyle='--')
337
        plt.axvline(x=mass_z0_fit, color='r', linestyle='--')
338
        plt.xlabel('Z boson mass (mass_z0) [GeV/c^2]')
339
        plt.ylabel('Z boson width (partial_width_z0) [GeV]')
340
341
        plt.title('Chi-squared Contours')
        plt.legend()
342
        plt.savefig("chi_square_contours.png", format='png')
343
        plt.show()
344
345
    def plot_chi_square_3d_contours(center_of_mass_energy,
347
        cross_section,
                                      uncertainty, mass_z0_range,
348
                                      partial_width_z0_range):
349
350
        (This function is for No.3 additional feature)
351
352
        Plot a 3D contour of the chi-square values, which is helpful in
353
             more vivid
        visualisasion for chisquare contour. This is an optional output
354
            , which is
355
        determined by user.
356
        Args:
357
        center_of_mass_energy (array): Array of center-of-mass energies
358
        cross_section (array): Array of measured cross sections.
359
        uncertainty (array): Array of uncertainties in the cross
360
361
        measurements.
        mass_z0_range (tuple): Range for the ZO boson mass
362
363
        range.
364
        partial_width_z0_range (tuple): Range for the Z0
        boson width range.
365
        mass_z0_fit (float): Fitted value of Z0 boson mass.
366
367
        partial_width_z0_fit (float): Fitted value of Z0 boson width.
368
        Returns:
369
370
        None
371
        mass_z0_values = np.linspace(*mass_z0_range, 100)
372
        partial_width_z0_values = np.linspace(*partial_width_z0_range,
373
            100)
374
        mass_z0_grid, partial_width_z0_grid = np.meshgrid(
            mass_z0_values, partial_width_z0_values)
375
        chi_sq_grid = np.zeros_like(mass_z0_grid)
376
377
        # Calculate chi-square values
        for i in range(mass_z0_grid.shape[0]):
379
             for j in range(mass_z0_grid.shape[1]):
380
381
                 chi_sq_grid[i, j] = chi_square([mass_z0_grid[i, j],
                                                   partial_width_z0_grid[i
382
                                                       , j]],
                                                  center_of_mass_energy,
383
                                                  cross_section,
384
```

```
uncertainty)
        fig = plt.figure()
385
        axes = fig.add_subplot(111, projection='3d')
386
387
        # Plot a 3D contour
388
        axes.contour3D(mass_z0_grid, partial_width_z0_grid,
389
390
                        chi_sq_grid, 50, cmap='viridis')
391
        # Label the axes
392
        axes.set_xlabel('Z boson mass (mass_z0) [GeV/c^2]')
393
        axes.set_ylabel('Z boson width (partial_width_z0) [GeV]')
394
        axes.set_zlabel('Chi-squared')
395
        axes.legend()
396
        # Save the plot in png version
397
        plt.savefig("3D_chi_square_contours.png", format='png')
398
        plt.show()
399
400
401
    def find_uncertainty_hill_climbing(data, best_fit_parameters,
402
                                          chi_squared_min, param_index,
403
                                          delta_chi_square=1.0):
405
        Estimate the uncertainty of the fit parameters using the hill-
406
            climbing
        method.
407
408
409
        Args:
        data (array): data array.
410
        best_fit_parameters (list): List of the best fit parameters.
411
        chi_squared_min (float): The minimum chi-square.
412
        param_index (int): Index of the parameter to calculate the
413
            uncertainty.
        step_size (float): The step size
414
        delta_chi_sq (float): The change in chi-square value.
415
416
417
        float: The uncertainties of two parameters.
418
419
        # Set proper step size
        step_size = 0.00001
420
421
        parameter_plus = best_fit_parameters[param_index]
        parameter_minus = best_fit_parameters[param_index]
422
        chi_squared_plus = chi_squared_min
423
        chi_squared_minus = chi_squared_min
424
        # Apply hill climbing method from two directions
425
        while chi_squared_plus - chi_squared_min < delta_chi_square:</pre>
426
            parameter_plus += step_size
427
            parameters = best_fit_parameters.copy()
428
             parameters[param_index] = parameter_plus
429
             chi_squared_plus = chi_square(
430
                 parameters, data[:, 0], data[:, 1], data[:, 2])
431
432
        while chi_squared_minus - chi_squared_min < delta_chi_square:</pre>
433
434
             parameter_minus -= step_size
             parameters = best_fit_parameters.copy()
435
436
             parameters[param_index] = parameter_minus
             chi_squared_minus = chi_square(
437
438
                 parameters, data[:, 0], data[:, 1], data[:, 2])
```

```
uncertainty = (parameter_plus - parameter_minus) / 2
439
440
        return uncertainty
441
442
    def calculate_lifetime(partial_width_z0):
443
444
        Calculate the lifetime of the ZO boson.
445
446
447
        partial_width_z0 (float): The partial width of the ZO boson.
448
449
450
        Returns:
        float: The calculated lifetime of the ZO boson in seconds.
451
452
        \mbox{\tt\#} Get the lifetime value for ZO
453
        lifetime_z0 = HBAR_GEV_S / partial_width_z0
454
455
        return lifetime_z0
456
457
    def calculate_lifetime_uncertainty(partial_width_z0,
458
                                        partial_width_uncertainty):
460
        Calculate the uncertainty in the lifetime of the ZO boson.
461
462
463
        Args:
        partial_width_z0 (float): The partial width of the ZO boson in
464
           GeV.
        partial_width_uncertainty (float): The uncertainty in the
465
            partial width.
466
        Returns:
467
        float: The uncertainty in the lifetime of the ZO boson in
468
469
        lifetime_uncertainty = HBAR_GEV_S / (partial_width_z0 ** 2) * \
470
471
            partial_width_uncertainty
        return lifetime_uncertainty
472
473
474
475
    def approximate_ellipse_center(center_of_mass_energy, cross_section
                                    uncertainty, mass_z0_range,
476
477
                                    partial_width_z0_range):
478
        (This function is for No.4 additional feature)
479
        Approximate the geometric center of the chi-square ellipse for
480
        sigma level. It is really helpful for further anlysis of the
481
            quality of the
        fit.
483
        Args:
484
        485
        mass_z0_range, partial_width_z0_range: Ranges for Z0 boson mass
486
             and width.
487
488
        Returns:
```

```
tuple: Approximate geometric center coordinates of the chi-
489
            square ellipse.
490
        sigma_level = 1
491
        mass_z0_values = np.linspace(*mass_z0_range, 100)
492
        partial_width_z0_values = np.linspace(*partial_width_z0_range,
493
            100)
        mass_z0_grid, partial_width_z0_grid = np.meshgrid(
494
             mass_z0_values, partial_width_z0_values)
         chi_sq_grid = np.zeros_like(mass_z0_grid)
496
497
        # Calculate chi-square values
498
        for i in range(mass_z0_grid.shape[0]):
499
             for j in range(mass_z0_grid.shape[1]):
                 chi_sq_grid[i, j] = chi_square(
501
                     [mass_z0_grid[i, j], partial_width_z0_grid[i, j]],
502
503
                     center_of_mass_energy, cross_section, uncertainty)
504
        min_chi_sq = chi_sq_grid.min()
505
        chi_sq_threshold = min_chi_sq + sigma_level**2
506
        # Find points within the sigma level contour
508
        within_contour = np.where(chi_sq_grid <= chi_sq_threshold)</pre>
509
        {\tt contour\_mass\_z0 = mass\_z0\_grid[within\_contour]}
510
        \verb|contour_partial_width_z0| = \verb|partial_width_z0_grid[within_contour||
511
512
        # Calculate the approximate geometric center
513
        center_mass_z0 = np.mean(contour_mass_z0)
514
        center_partial_width_z0 = np.mean(contour_partial_width_z0)
515
        return center_mass_z0, center_partial_width_z0
517
518
519
    def gaussian(x_value, mean_value, sigma):
          ""Returns the value of a Gaussian probability density function
520
             at x."""
        return 1 / (sigma * np.sqrt(2 * np.pi)) * \
522
            np.exp(-0.5 * ((x_value - mean_value) / sigma) ** 2)
523
524
    def plot_gaussian_distribution(x_value, mean_value,
        x_value_uncertainty,
                                     title, filename):
526
         (This function is for No.5 additional feature)
528
529
        Plots a Gaussian distribution for a given parameter.
530
531
        x_values = np.linspace(x_value - 10 * x_value_uncertainty,
            x_value + 10 *
                                 x_value_uncertainty, 1000)
534
        y_values = gaussian(x_values, mean_value, x_value_uncertainty)
536
537
        plt.figure(figsize=(12, 6))
        plt.plot(x_values, y_values, label=f'Gaussian of {title}')
538
539
        plt.xlabel('Values')
```

```
plt.ylabel('Probability Density')
540
541
        plt.title(f'Distribution for ZO Boson {title}')
        plt.legend()
542
        plt.grid(True)
        plt.savefig(filename)
544
        plt.show()
545
546
547
    def save_data_to_csv_file(filename, data):
548
         (This function is for No.6 additional feature)
550
        Save the final filtered data as a CSV file for future use. This
552
             is an
        optional output determined by user.
554
        Args:
        filename (str): Name of the CSV file to be saved.
556
557
        data (list of tuples): Data to be saved
558
        with open(filename, mode='w', newline='', encoding='utf-8') as
            file:
             # Write the header
560
            file.write("Time (s),Fractional Intensity\n")
561
             # Write the data rows
562
563
            for time, frac_intensity in data:
                 file.write(f"{time},{frac_intensity}\n")
564
565
566
    def assess_fit_quality(fitted_mass_z0, fitted_width_z0,
567
        center_mass_z0,
                             center_partial_width_z0):
568
569
        Assess the quality of the fit by checking if the Gaussian
570
            center points
        for mass_z0 and width_z0 are within the uncertainty range of
            the fitted
572
        values.
573
574
        Args:
        fitted_mass_z0 (float): Fitted Z boson mass.
575
        fitted_width_z0 (float): Fitted Z boson width.
576
         center_mass_z0 (float): Center point of the Gaussian for
577
            mass z0.
         center_width_z0 (float): Center point of the Gaussian for
578
            width_z0.
579
580
        Returns:
        None
581
582
        tolerance = 0.0001
583
        mass_in_good_quality = abs(
584
585
            fitted_mass_z0 - center_mass_z0) <= tolerance</pre>
        width_in_good_quality = abs(
586
587
            fitted_width_z0 - center_partial_width_z0) <= tolerance</pre>
588
589
        if mass_in_good_quality and width_in_good_quality:
```

```
print("Fit Quality: Good. Best fit is off-center but within
590
                  tolerance,'
                   "indicating the distributions are Gaussian
591
                        distribution.")
592
        else:
             print("Fit Quality: Not Good. Best fit point deviates from
594
                 the"
                   "ellipse's center. the output distribution plots are
                       not"
                   "gaussion distribution, Consider re-optimizing chi-
596
                        square.")
597
598
    def main():
599
600
601
        Main function
602
603
        # Combine two data sets
604
         combined_data = read_and_combine_data(FILE_NAME_1, FILE_NAME_2)
605
        if combined_data is None:
606
             return
607
608
        filtered_data_1 = filter_data(combined_data, num_std=3)
609
        # Get the data arrays filtered by filter_1 function
610
        center_of_mass_energy = filtered_data_1[:, 0]
611
        cross_section = filtered_data_1[:, 1]
612
        uncertainty = filtered_data_1[:, 2]
613
614
615
        # Take the initial guess for two parameters
        initial_guesses = [90, 3]
616
        result = min_chi_square(
617
618
             center_of_mass_energy, cross_section, uncertainty,
                 initial_guesses)
619
        mass_z0_fit, partial_width_z0_fit = result[0]
620
621
        # Get the final data arrays which filtered by filter_1 &
            filter_2 functions
         final_filter_data = filter_data_2(
622
             {\tt center\_of\_mass\_energy}\;,\;\;{\tt cross\_section}\;,\;\;{\tt mass\_z0\_fit}\;,
623
             partial_width_z0_fit, filtered_data_1)
624
         center_of_mass_energy_2 = final_filter_data[:, 0]
625
         cross_section_2 = final_filter_data[:, 1]
626
        uncertainty_2 = final_filter_data[:, 2]
627
628
        result_2 = min_chi_square(
629
630
             center_of_mass_energy_2, cross_section_2, uncertainty_2,
             initial_guesses)
631
        mass_z0_fit, partial_width_z0_fit = result_2[0]
632
633
        # Calculate the reduced chi_square
634
635
        num_data_points = len(cross_section_2)
        num_parameters = len(result_2[0])
636
637
        reduced_chi_sq_2 = calculate_reduced_chi_square(
             result_2[1], num_data_points, num_parameters)
638
639
```

```
# Plot the min_chi_square fit with scatted datas
640
         plot_data_and_fit(center_of_mass_energy_2,
641
                            cross_section_2, uncertainty_2, mass_z0_fit,
642
                            partial_width_z0_fit)
643
644
         # Plot the chi_square contours
645
         mass_z0_range = [mass_z0_fit - 0.03, mass_z0_fit + 0.03]
         partial_width_z0_range = [
647
             partial_width_z0_fit - 0.03, partial_width_z0_fit + 0.03]
648
649
         plot_chi_square_contours(
650
             center_of_mass_energy_2, cross_section_2, uncertainty_2,
651
             mass_z0_range, partial_width_z0_range, mass_z0_fit,
             partial_width_z0_fit)
652
653
         \mbox{\tt\#} Get the result of Z boson mass, Z boson width and minimum
654
             chi_square
655
        {\tt mass\_z0\_fit}\;,\;\;{\tt partial\_width\_z0\_fit}\;\;{\tt =}\;\;{\tt result\_2}\,[{\tt 0}]
         chi_squared_min = result_2[1]
656
657
         \hbox{\tt\# Calculate uncertainties for $\tt mass\_z0$, $\tt partial\_width\_z0$}
658
         mass_z0_uncertainty = find_uncertainty_hill_climbing(
             final_filter_data, result_2[0], chi_squared_min,
660
                 param_index=0)
         partial_width_z0_uncertainty = find_uncertainty_hill_climbing(
661
             final_filter_data, result_2[0], chi_squared_min,
662
                 param_index=1)
         # Calculate the lifetime_z0
663
         lifetime_z0 = calculate_lifetime(partial_width_z0_fit)
664
         # Cal
665
         lifetime_z0_uncertainty = calculate_lifetime_uncertainty(
666
             partial_width_z0_fit, partial_width_z0_uncertainty)
667
668
         \# Plot the Gaussian distributions for mass_z0 and
669
             partial_width_z0
670
         center_mass_z0, center_partial_width_z0 =
             approximate_ellipse_center(
             center_of_mass_energy_2, cross_section_2, uncertainty_2,
671
672
             mass_z0_range, partial_width_z0_range)
673
674
         plot_gaussian_distribution(mass_z0_fit, center_mass_z0,
                                      mass_z0_uncertainty, 'Mass (mass_z0)
675
676
                                       'gaussian_distribution_mass_z0.png')
677
         plot_gaussian_distribution(partial_width_z0_fit,
678
             center_partial_width_z0,
                                      partial_width_z0_uncertainty,
679
                                       'Partial Width (partial_width_z0)',
680
681
                                           gaussian_distribution_partial_width_z0
                                           .png')
682
683
         # Check the quality of the fit with the gaussian plots
         assess_fit_quality(mass_z0_fit, partial_width_z0_fit,
684
             center_mass_z0,
                              center_partial_width_z0)
685
686
```

```
# Print result values and their uncertainties
687
        print(
            f"Z boson mass (mass_z0): "
689
            f"{mass_z0_fit:.4g}
                                  {mass_z0_uncertainty:.5f} GeV/c^2"
690
691
        print(
692
            f"Z boson width (partial_width_z0): "
693
            f"{partial_width_z0_fit:.4g}
694
                 partial_width_z0_uncertainty:.5f} GeV"
695
        print(
696
            f"Z boson lifetime: "
697
            f"{lifetime_z0:.2e} s
                                       {lifetime_z0_uncertainty:.2e} s"
698
699
        print(f"Minimum_value chi-squared: {result_2[1]:.3f}")
700
        print(f"Reduced chi_squared: {reduced_chi_sq_2:.3f}")
701
702
        # Interactive menu
703
704
        while True:
            user_input = input(
705
706
                 "Enter '1' for 3D Chi-Square Contours, "
                 "'2' to save filtered data, or 'quit' to exit: "
707
708
709
             if user_input.lower() == 'quit':
710
                 print("Exiting the program.")
711
                 break
712
713
             if user_input == '1':
714
                 # Plot chi-square contours with contour lines
715
716
                 plot_chi_square_3d_contours(center_of_mass_energy_2,
                                               cross_section_2.
717
                                                   uncertainty_2,
718
                                               mass_z0_range,
                                                   partial_width_z0_range)
719
             elif user_input == '2':
                 # Save filtered data
720
721
                 filename_data = 'final_filtered_data.csv'
                 np.savetxt(filename_data, final_filter_data, delimiter=
722
                             header='Center_of_Mass_Energy, Cross_Section
723
724
                                 Uncertainty',
                             comments='', fmt='%f')
                 print(f"Filtered data saved to {filename_data}.")
726
727
728
    if __name__ == "__main__":
729
        main()
730
```