Physics 474 - Spring 2023 Homework 1

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In this homework we will practice fitting a function with parameters to some data.

skills we will excercise:

- · reading in data
- plotting data
- · writing user defined functions
- fitting a function to data with 'curve_fit'
- calculating χ^2 and χ^2 probaility
- plotting residuals
- analyzing data and making observations

The data we will be using is global temperature data compiled from The current citation for this dataset is:

```
Rohde, R. A. and Hausfather, Z.: The Berkeley Earth Land/Ocean Temperature Record, Earth Syst. Sci. Data, 12, 3469@3479, https://doi.org/10.5194/essd-12-3469-2020, 2020.
```

Data is: year, month, delta_T(C), T_error(C)

where delta_T = Temp - (Jan 1951-Dec 1980 global mean temperature)

Estimated Jan 1951-Dec 1980 global mean temperature (C) Using air temperature above sea ice: 14.105 +/- 0.022

The data is provided in a comma-separated-value file named 'global_temps_datafile.csv'

Note that the temperature data is provided as a ΔT from the global mean 1951-1980 temperature of

$$T_{mean}^{51-80} = 14.105 \pm 0.022^{\circ} C$$

Part 1) (1 pt)

Read in the data file and print the shape of the file

```
In []: #Your code here...
import pandas as pd
```

```
temp_data = pd.read_csv('./global_temps_datafile_worker.csv', sep=',', engine=
print(temp_data.shape)
(2074, 4)
```

Part 2) (4 pts)

Plot

- the data points with errorbars vs year.
- ullet a horizontal line at $T_{mean}^{51-80}=14.105^{\circ}C$
- suggestion: use "alpha=0.2" in plotting data points

About errors: There are three errors that need to be added in quadrature

- the error on each of the measurements σ_T from the file
- ullet error on the T_{mean}^{51-80} of $\sigma_{T_{mean}}=0.022$
- ullet a random systematic error of $\sigma_{sus}=0.13$

Recall from lecture then

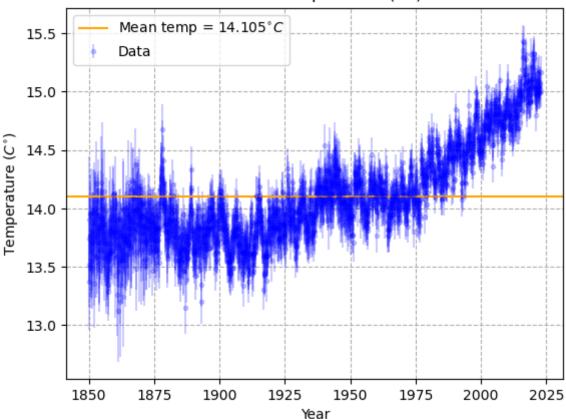
$$\sigma_{tot}^2 = \sigma_T^2 + \sigma_{T_{mean}}^2 + \sigma_{sys}^2$$

```
import matplotlib.pyplot as plt

temp_data_cp = temp_data.copy()
  temp_data_cp['date'] = temp_data_cp['year'] + (temp_data_cp['month'] * (1/12))
  temp_data_cp['error'] = ((temp_data_cp['error_t'])**2 + (0.022)**2 + (0.13)**2)
  temp_data_cp['temp'] = 14.105 + temp_data_cp['delta_t']

plt.errorbar(temp_data_cp['date'],temp_data_cp['temp'],yerr=temp_data_cp['error_plt.axhline(y=14.105, color='orange', label=r'Mean_temp = $14.105^{\circ} C$')
  plt.grid(True,linestyle='--')
  plt.title(r'Year_vs_Temperature ($C^{\circ}$)')
  plt.xlabel('Year')
  plt.ylabel(r'Temperature ($C^{\circ}$)')
  plt.legend()
  plt.show()
```





Part 2 Observations:

Part 3a) (2 pts)

Fit a straight-line (i.e. first-order polynomial) to the data using 'curve_fit' from some point starting after the pre-industrial era (i.e. sometime after 1900). You might try after 1900, 1925, 1950, 1980...

sometime after 1900 but before the visible rise around 1980.

```
In []: #Your code here...
import matplotlib.pyplot as plt
from scipy.optimize import curve_fit
import math

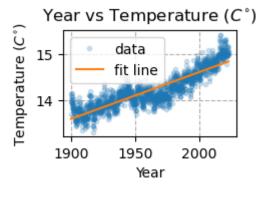
def straightLine(m, x, b):
    return (m*x) + b

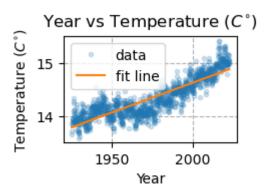
temp_data_cp = temp_data.copy()
temp_data_cp['date'] = temp_data_cp['year'] + (temp_data_cp['month'] * (1/12))
temp_data_cp['error'] = ((temp_data_cp['error_t'])**2 + (0.022)**2 + (0.13)**2)
temp_data_cp['temp'] = 14.105 + temp_data_cp['delta_t']

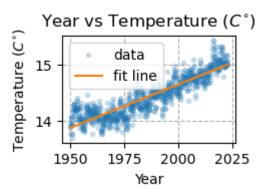
fig, axs = plt.subplots(2,2)
```

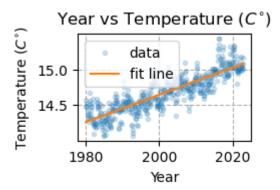
```
fig.tight layout(pad=5.0)
nineteen hundred = temp data cp[600:]
param1900, param1900_cov = curve_fit(straightLine, nineteen_hundred['date'], ni
nineteen hundred fit = straightLine(nineteen hundred['date'], param1900[0], par
nineteen_twentyfive = temp_data_cp[900:]
param1925, param1925_cov = curve_fit(straightLine, nineteen twentyfive['date'],
nineteen_twentyfive_fit = straightLine(nineteen_twentyfive['date'], param1925[(
nineteen fifty = temp data cp[1200:]
param1950, param1950_cov = curve_fit(straightLine, nineteen_fifty['date'], nineteen_fifty['date']
nineteen_fifty_fit = straightLine(nineteen_fifty['date'], param1950[0], param19
nineteen_eighty = temp_data_cp[1560:]
param1980, param1980 cov = curve fit(straightLine, nineteen eighty['date'], nir
nineteen_eighty_fit = straightLine(nineteen_eighty['date'], param1980[0], param
axs[0,0].plot(nineteen_hundred['date'], nineteen_hundred['temp'], '.', alpha=0.
axs[0,0].plot(nineteen_hundred['date'], nineteen_hundred_fit, '-', label='fit ]
axs[0,0].grid(True,linestyle='--')
axs[0,0].set xlabel('Year')
axs[0,0].set ylabel(r'Temperature ($C^{\circ}$)')
axs[0,0].legend()
axs[0,0].set_title(r'Year vs Temperature ($C^{\circ}$)')
axs[0,1].plot(nineteen twentyfive['date'], nineteen twentyfive['temp'], '.', a]
axs[0,1].plot(nineteen twentyfive['date'], nineteen twentyfive fit, '-', label=
axs[0,1].grid(True,linestyle='--')
axs[0,1].set xlabel('Year')
axs[0,1].set ylabel(r'Temperature ($C^{\circ}$)')
axs[0,1].legend()
axs[0,1].set title(r'Year vs Temperature ($C^{\circ}$)')
axs[1,0].plot(nineteen fifty['date'], nineteen fifty['temp'], '.', alpha=0.2, ]
axs[1,0].plot(nineteen fifty['date'], nineteen fifty fit, '-', label='fit line'
axs[1,0].grid(True,linestyle='--')
axs[1,0].set xlabel('Year')
axs[1,0].set ylabel(r'Temperature ($C^{\circ}$)')
axs[1,0].legend()
axs[1,0].set title(r'Year vs Temperature ($C^{\circ}$)')
axs[1,1].plot(nineteen eighty['date'], nineteen eighty['temp'], '.', alpha=0.2,
axs[1,1].plot(nineteen eighty['date'], nineteen eighty fit, '-', label='fit lir
axs[1,1].grid(True,linestyle='--')
axs[1,1].set xlabel('Year')
axs[1,1].set_ylabel(r'Temperature ($C^{\circ}$)')
axs[1,1].legend()
axs[1,1].set title(r'Year vs Temperature ($C^{\circ}$)')
# print(f'slope: {round(param1900[0],4)} +- {round(math.sqrt(param1900 cov[0,0]
# print(f'intercept: {round(param1900[1],1)} +- {round(math.sqrt(param1900 cov
```

Out[]: Text(0.5, 1.0, 'Year vs Temperature (\$C^{\\circ}\$)')









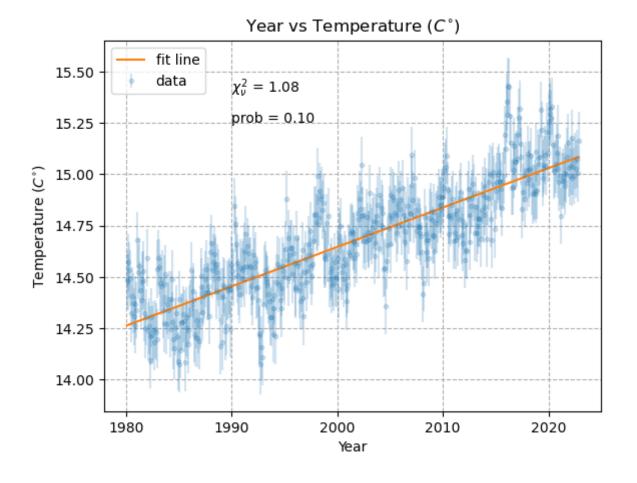
Part 3b) (3 pts)

- 1) Calculate and output the χ^2 , dof, and χ^2 probability for the fit
- 2) Plot
 - the data with errorbars for the entire range of years
 - the best fit line over the years for the fit
 - put labels on the data, fit, axes, etc...

```
In []:
        #Your code here...
        import matplotlib.pyplot as plt
        from scipy.optimize import curve_fit
        import scipy.stats as st
        import math
        import numpy as np
        def straightLine(m, x, b):
            return (m*x) + b
        def chi squared(Theory, Data, sigma):
             if np.size(Theory)==np.size(Data) and np.size(Data)==np.size(sigma):
                chi2=np.sum((Theory-Data)**2/sigma**2)
                return chi2
            else:
                print('error - arrays of unequal size')
                return -1.
        temp data cp = temp data.copy()
        temp_data_cp['date'] = temp_data_cp['year'] + (temp_data_cp['month'] * (1/12))
```

```
temp_data_cp['error'] = ((temp_data_cp['error_t'])**2 + (0.022)**2 + (0.13)**2)
temp data cp['temp'] = 14.105 + temp data cp['delta t']
nineteen_eighty = temp_data_cp[1560:]
param1980, param1980_cov = curve_fit(straightLine, nineteen_eighty['date'], nir
nineteen eighty fit = straightLine(nineteen eighty['date'], param1980[0], param
chi2 = chi_squared(nineteen_eighty_fit, nineteen_eighty['temp'], nineteen_eight
dof = nineteen eighty.shape[0] - 2
prob = st.chi2.sf(chi2,dof)
plt.errorbar(nineteen_eighty['date'], nineteen_eighty['temp'], yerr=nineteen_eighty['date']
plt.plot(nineteen_eighty['date'], nineteen_eighty_fit, '-', label='fit line')
plt.grid(True, linestyle='--')
plt.xlabel('Year')
plt.ylabel(r'Temperature ($C^{\circ}$)')
plt.legend()
plt.text(1990,15.4,r'\chi_\nu^2\ = \{:.2f}'.format(chi2/dof))
plt.text(1990,15.25,r'prob = {:.2f}'.format(prob))
plt.title(r'Year vs Temperature ($C^{\circ}$)')
```

Out[]: Text(0.5, 1.0, 'Year vs Temperature (\$C^{\\circ}\$)')

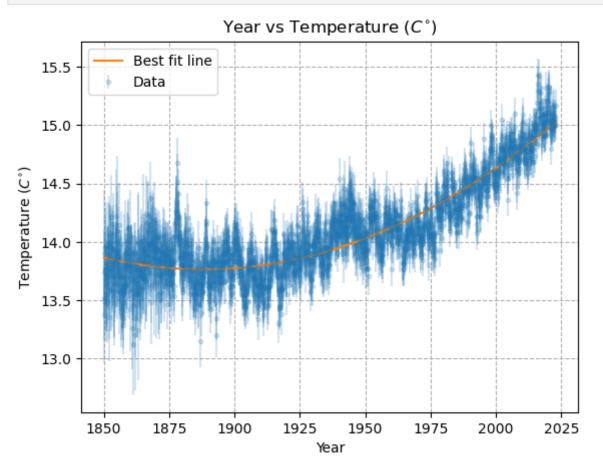


Part 3 Observations: (will depend on range used for linear fit)

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Fit a Quadratic (i.e. second-order polynomial) to the data using 'curve_fit' ffor the entire range of years

```
In []:
        #Your code here...
        import matplotlib.pyplot as plt
        from scipy.optimize import curve fit
        def objective(x, a, b, c):
            return a * x + b * x**2 + c
        temp_data_cp = temp_data.copy()
        temp_data_cp['date'] = temp_data_cp['year'] + (temp_data_cp['month'] * (1/12))
        temp_data_cp['error'] = ((temp_data_cp['error_t'])**2 + (0.022)**2 + (0.13)**2)
        temp_data_cp['temp'] = 14.105 + temp_data_cp['delta_t']
        param, param_cov = curve_fit(objective, temp_data_cp['date'], temp_data_cp['tem
        fit = objective(temp_data_cp['date'], param[0], param[1], param[2])
        plt.errorbar(temp_data_cp['date'],temp_data_cp['temp'],yerr=temp_data_cp['error
        plt.plot(temp_data_cp['date'], fit, '-', label='Best fit line')
        plt.grid(True, linestyle='--')
        plt.title(r'Year vs Temperature ($C^{\circ}$)')
        plt.xlabel('Year')
        plt.ylabel(r'Temperature ($C^{\circ}$)')
        plt.legend()
        plt.show()
        plt.show()
```



Part 4b) (3 pts)

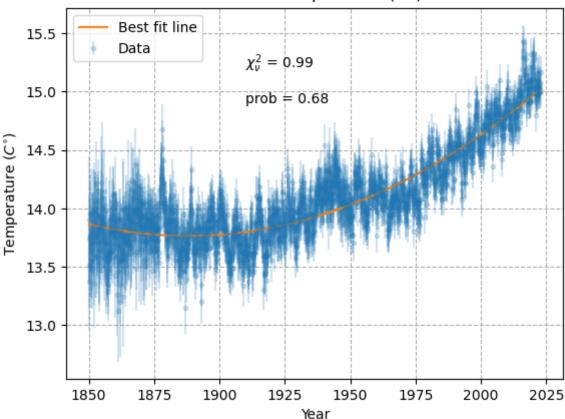
1) Calculate and output the χ^2 , dof, and χ^2 probability for the fit

- 2) Plot
 - the data with errorbars for the entire range of years
 - the best fit quadratic over the entire range of years
 - put labels on the data, fit, axes, etc...

```
In [ ]: #Your code here...
        import numpy as np
        import matplotlib.pyplot as plt
        import scipy.stats as st
                                   #for chi-squared probability
        from scipy.optimize import curve fit #for curve fit routine
        def chi_squared(Theory, Data, sigma):
            if np.size(Theory)==np.size(Data) and np.size(Data)==np.size(sigma):
                chi2=np.sum((Theory-Data)**2/sigma**2)
                return chi2
            else:
                print('error - arrays of unequal size')
                return -1.
        def objective(x, a, b, c):
            return a * x + b * x**2 + c
        temp data cp = temp data.copy()
        temp data cp['date'] = temp data cp['year'] + (temp data cp['month'] * (1/12))
        temp data cp['error'] = ((temp data cp['error t'])**2 + (0.022)**2 + (0.13)**2)
        temp data cp['temp'] = 14.105 + temp data cp['delta t']
        param, param cov = curve fit(objective, temp data cp['date'], temp data cp['tem
        fit = objective(temp data cp['date'], param[0], param[1], param[2])
        chi2 = chi squared(objective(temp data cp['date'], param[0], param[1], param[2]
        dof = temp data cp.shape[0] - 3
        prob = st.chi2.sf(chi2,dof)
        plt.figure()
        plt.errorbar(temp_data_cp['date'],temp_data_cp['temp'],yerr=temp_data_cp['error
        plt.plot(temp data cp['date'], fit, '-', label='Best fit line')
        plt.grid(True,linestyle='--')
        plt.title(r'Year vs Temperature ($C^{\circ}$)')
        plt.xlabel('Year')
        plt.ylabel(r'Temperature ($C^{\circ}$)')
        plt.legend()
        plt.text(1910,15.2,r'$\chi_\nu^2$ = {:.2f}'.format(chi2/dof))
        plt.text(1910,14.9,r'prob = {:.2f}'.format(prob))
        plt.show()
```

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Part 4 Observations:

Part 5: (2 pts)

Make 2 subplots (2 rows x 1 column)

- top: residuals with errorbars for the straight-line fit for years used in the fit
- bottom: residuals with errorbars for quadratic fit for years used in the fit

Use the same limits on the x-axis for both 1850-2025

```
In []: #Your code here...
    from numpy import random, mean, std
    import matplotlib.pyplot as plt
    from scipy.optimize import curve_fit #for curve_fit routine

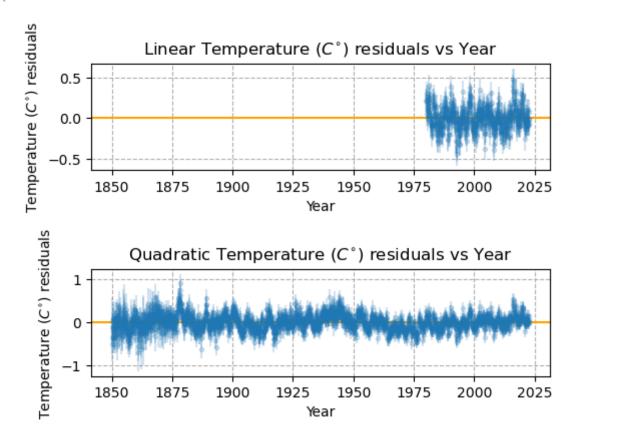
def objectiveStraight(m, x, b):
        return (m*x) + b

def objectiveCurve(x, a, b, c):
        return a * x + b * x**2 + c

temp_data_cp = temp_data.copy()
    temp_data_cp['date'] = temp_data_cp['year'] + (temp_data_cp['month'] * (1/12))
    temp_data_cp['error'] = ((temp_data_cp['error_t'])**2 + (0.022)**2 + (0.13)**2)
```

```
temp data cp['temp'] = 14.105 + temp data cp['delta t']
nineteen_eighty = temp_data_cp[1560:]
paramStraight, paramStraight_cov = curve_fit(straightLine, nineteen_eighty['dat'
paramCurve, paramCurve cov = curve fit(objectiveCurve, temp data cp['date'], te
fig, axs = plt.subplots(2)
fig.tight_layout(pad=5.0)
axs[0].axhline(y=0, color='orange')
axs[0].errorbar(temp_data_cp['date'], nineteen_eighty['temp']-objectiveStraight
axs[0].grid(True,linestyle='--')
axs[0].set_xlabel('Year')
axs[0].set ylabel(r'Temperature ($C^{\circ}$) residuals')
axs[0].set_title(r'Linear Temperature ($C^{\circ}$) residuals vs Year')
axs[1].axhline(y=0, color='orange')
axs[1].errorbar(temp_data_cp['date'], temp_data_cp['temp']-objectiveCurve(temp
axs[1].grid(True, linestyle='--')
axs[1].set_xlabel('Year')
axs[1].set_ylabel(r'Temperature ($C^{\circ}$) residuals')
axs[1].set_title(r'Quadratic Temperature ($C^{\circ}$) residuals vs Year')
```

Out[]: Text(0.5, 1.0, 'Quadratic Temperature (\$C^{\\circ}\$) residuals vs Year')



Part 5 observations: (will depend on range used for linear fit)

Part 6: (3 pts)

On a single plot show

- data with errorbars
- both fits projected out 100 years
- a horizontal line from 1850-1900 with the mean pre-industrial temperature
- ullet a horizontal line at the 1951-1980 T_{mean}
- ullet a horizontal line at $\Delta T=2^{\circ}C$ above the pre-industrial temperature

put all appropriate labels, etc... on plot

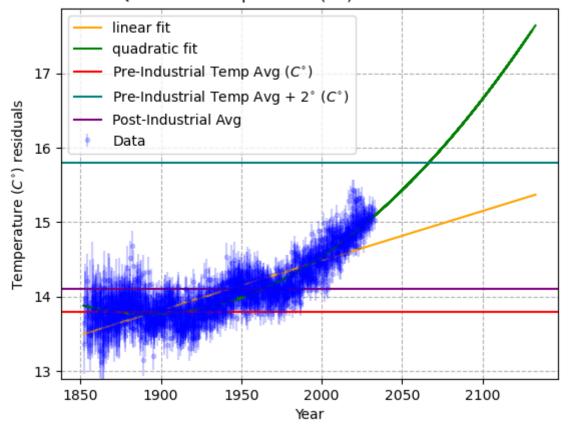
```
In [ ]: #Your code here...
        import matplotlib.pyplot as plt
        import scipy.stats as st
        from scipy.optimize import curve fit
        from numpy import mean
        import pandas as pd
        def objectiveStraight(m, x, b):
            return (m*x) + b
        def objectiveCurve(x, a, b, c):
            return a * x + b * x**2 + c
        temp data cp = temp data.copy()
        temp data cp['date'] = temp data cp['year'] + temp data cp['month']
        temp data cp['error'] = ((temp data cp['error t'])**2 + (0.022)**2 + (0.13)**2)
        temp data cp['temp'] = 14.105 + temp data cp['delta t']
        projection data = []
        for i in range(1850,2122,1):
            for j in range(1,13,1):
                if i == 1850 and j == 1:
                    continue
                else:
                    data = [i,j]
                    projection data.append(data)
        projection = pd.DataFrame(projection data, columns=['year', 'month'])
        projection['date'] = projection['year'] + projection['month']
        projection['error'] = ((temp data cp['error t'])**2 + (0.022)**2 + (0.13)**2)**
        projection['temp'] = 14.105 + temp data cp['delta t']
        paramStraight, paramStraight cov = curve fit(objectiveStraight, temp data cp['c
        straight fit = objectiveStraight(projection['date'], paramStraight[0], paramStr
        paramCurve, paramCurve cov = curve fit(objectiveCurve, temp data cp['date'], te
        quadratic fit = objectiveCurve(projection['date'], paramCurve[0], paramCurve[1]
        projection['linear fit'] = straight fit
        projection['quadratic fit'] = quadratic fit
        pre industrial data = temp data cp[:600] # 1850 - 1900
        pre industrial mean = mean(pre industrial data['temp'])
        post industrial data = temp data cp[1200:1560] # 1950 - 1980
```

```
post_industrial_mean = mean(post_industrial_data['temp'])

# print(projection)

plt.errorbar(projection['date'], projection['temp'], yerr=projection['error'], fmt
plt.plot(projection['date'], projection['linear_fit'], '-', color='orange', lat
plt.plot(projection['date'], projection['quadratic_fit'], '-', color='green', l
plt.axhline(y=pre_industrial_mean, color='red', label=r'Pre-Industrial Temp Avc
plt.axhline(y=pre_industrial_mean+2, color='teal', label=r'Pre-Industrial Temp
plt.axhline(y=post_industrial_mean, color='purple', label=r'Post-Industrial Avc
plt.grid(True,linestyle='--')
plt.ylabel(r'Temperature ($C^{\circ}) residuals')
plt.xlabel('Year')
plt.title(r'Quadratic Temperature ($C^{\circ}) residuals vs Year')
plt.legend()
plt.show()
```

Quadratic Temperature (C°) residuals vs Year



Final Observations and Summary:

Pretty interesting to see that the linear fit line does not cross above $2^{\circ}C$ before the hundred year mark. Unfortunately it seems we have experienced exponential growth since 1950. It would be interesting to see the population data in combination with this data, and any projections there are on what will happen when populations decrease.