

# Tutorial\_2\_part\_b

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## 1 Interpolation functions

### 1.1 Junior mentor:

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### 1.2 Objectives of the tutorial

1. Differentiate between interpolating functions and curve fitting
2. Create 1-D interpolations
3. Identify the different types of 1-D interpolating functions

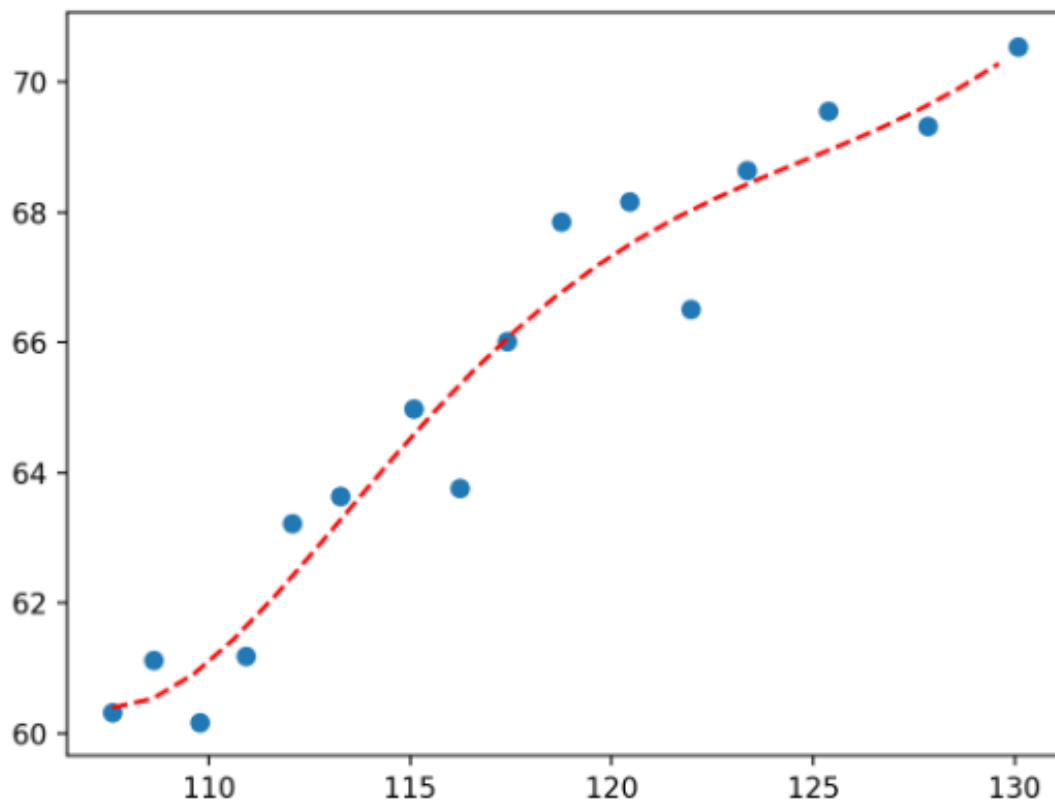
#### 1.2.1 Curve fitting

In Tutorial 2 part a, we learned that fitting a curve to a set of data means to find the curve that better describes most of the data points. However, the fitted curve does not pass through all of the data points.

```
[1]: from IPython import display
```

```
[2]: display.Image("./fit.png")
```

```
[2]:
```

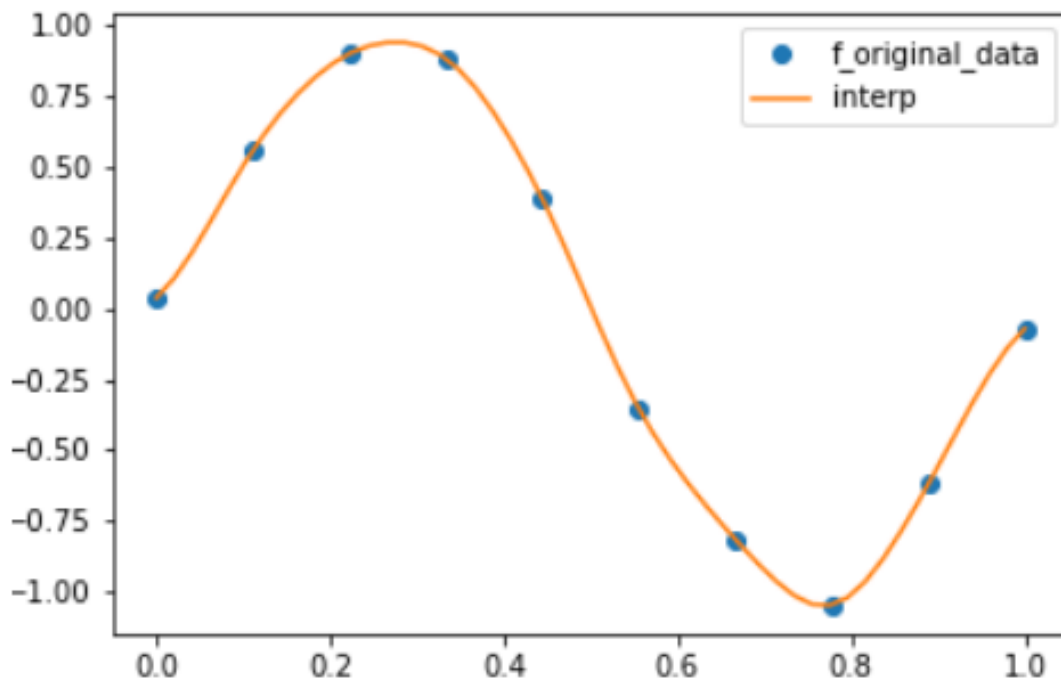


### 1.2.2 Interpolation

Given a set of data points  $(x_i, f_i)$ , with  $x_{min} \leq x_i \leq x_{max}$ , an interpolation function  $f(x)$  is a curve that passes through all of the data points available in the data set. Once we have found the interpolation functions, we can evaluate it at different locations  $f(x')$ , with  $x_{min} \leq x' \leq x_{max}$ , i.e. it can only be evaluated in the interval  $[x_{min}, x_{max}]$

```
[3]: display.Image("./interp.png")
```

```
[3]:
```



## 2 Example 1: Basic 1-D interpolation

For this example we use the function `interp1d`, contained inside the `scipy.interpolate` library

```
[4]: import numpy as np # For maths
      from scipy.interpolate import interp1d # For 1-D interpolations
      import matplotlib.pyplot as plt # For plotting
```

### 2.0.1 1. Let us create some data points

```
[5]: x = np.linspace(0,10,15)
      y = np.exp(-x/2)
```

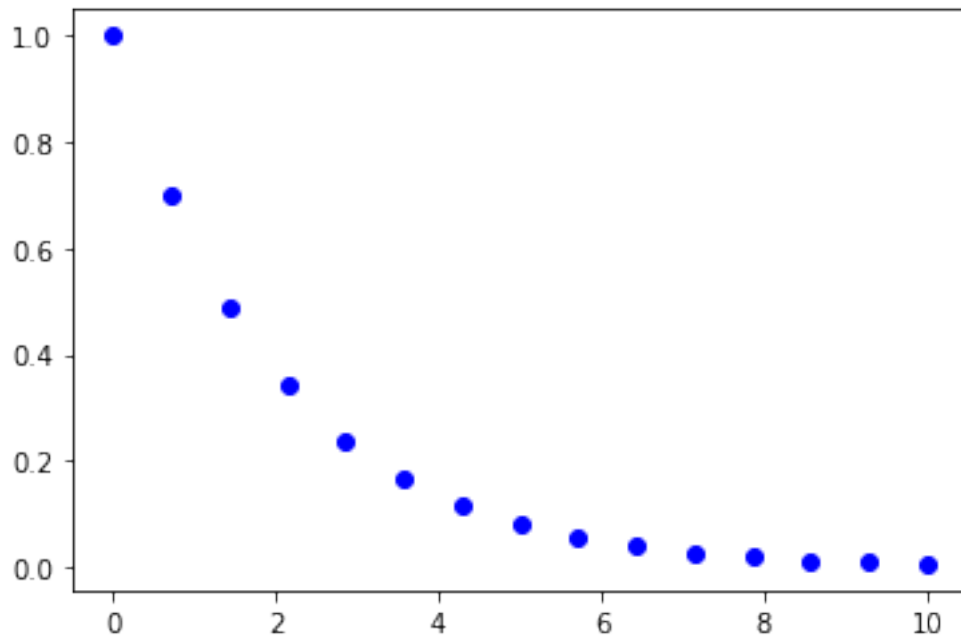
```
[6]: x
```

```
[6]: array([ 0.          ,  0.71428571,  1.42857143,  2.14285714,  2.85714286,
            3.57142857,  4.28571429,  5.          ,  5.71428571,  6.42857143,
            7.14285714,  7.85714286,  8.57142857,  9.28571429, 10.          ])
```

```
[7]: y
```

```
[7]: array([1.          , 0.69967254, 0.48954166, 0.34251886, 0.23965104,  
          0.16767725, 0.11731917, 0.082085   , 0.05743262, 0.04018403,  
          0.02811566, 0.01967175, 0.01376379, 0.00963014, 0.00673795])
```

```
[8]: plt.plot(x,y,"bo")  
     plt.show()
```



## 2.0.2 2. Do the interpolation

```
[9]: interp1d?
```

```
[10]: f = interp1d(x,y)
```

```
[11]: type(f)
```

```
[11]: scipy.interpolate.interpolate.interp1d
```

```
[12]: f(3.4)
```

```
[12]: array(0.18495096)
```

Let us generate a diferent set of x-values

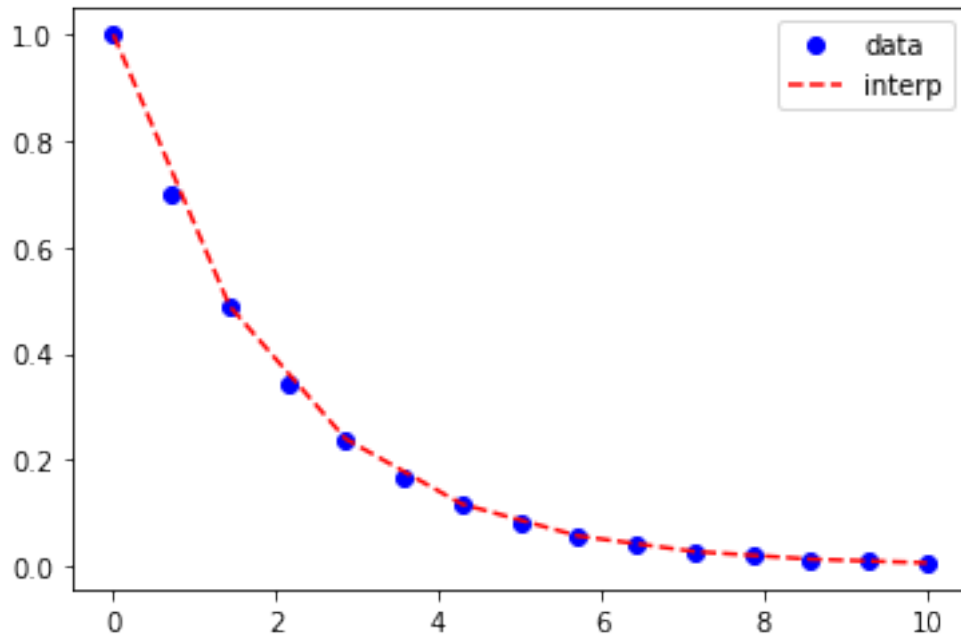
```
[13]: xnew = np.linspace(0,10, 8)
```

```
[14]: xnew
```

```
[14]: array([ 0.          ,  1.42857143,  2.85714286,  4.28571429,  5.71428571,
            7.14285714,  8.57142857, 10.          ])
```

Plot the interpolating function

```
[15]: plt.plot(x,y,'bo',label = "data")
plt.plot(xnew,f(xnew),'r--',label = "interp")
plt.legend(loc = "upper right")
plt.show()
```



```
[ ]: py.legend?
```

### 3 Types of 1-D interpolating functions

The default value for kind of interpolation is ``linear``

```
[17]: interp1d?
```

Let us see the effect of the different types of interpolating functions

#### 3.0.1 Create our data

```
[18]: x1 = np.linspace(0,10,11)
y1 = np.sin(x1)
x1new = np.linspace(0,10,100) #New data to evaluate after the interpolation is
↪ done
```

```
[19]: %matplotlib notebook
plt.plot(x1,y1,"bo",label = "data")
```

<IPython.core.display.Javascript object>

<IPython.core.display.HTML object>

```
[19]: [<matplotlib.lines.Line2D at 0x7f9a1ffb6390>]
```

```
[22]: %matplotlib notebook
plt.plot(x1,y1,"bo",label = "data")
# Let us now do the interpolation for all the different kinds
type_interp = ['linear', 'nearest', 'zero', 'slinear', 'quadratic', 'cubic']

for kind in type_interp:
    f1 = interp1d(x1,y1,kind = kind)
    ynew = f1(x1new)
    plt.plot(x1new,ynew,label = kind)
plt.legend(loc = 'lower right')
plt.show()
```

<IPython.core.display.Javascript object>

<IPython.core.display.HTML object>

## 4 Exercise

In this exercise we use interpolating functions to complete the tutorial on curve fitting from last week. Let us divide it into different steps

1. Prepare the data: Choose your preferred galaxy and load the RotationCurve data, as well as the RotationCurve\_baryons data. Create two data frames for the two different files
2. To have the data measured in the same `r_grid`, create an interpolating function for the stellar and gas velocities. (of course the data for radius to be used are the one given in the baryon file)
3. From the interpolating functions, compute the gas and stellar velocities, evaluated in the radius of the RotationCurve data frame
4. Add two new columns to the RotationCurve data frame, with the keywords ``stars circ velocity (km/s)'` and ``gas circ velocity (km/s)'`
5. Add a new column to the RotationCurve data frame, with the DM velocity,

which is computed by

$$v_{DM} = \sqrt{v_c^2 - v_{star}^2 - v_{gas}^2}.$$

Warning: If a given value of  $v_{DM}^2$  is negative, choose to change sign of  $v_{gas}^2$  for the specific point, as explained in Footnote 1 in readme .

6. Including errors: Add a new column to the RotationCurve data frame which includes a systematic error in the measurements. As explained in readme , the error is given by the 5 % of the last measured circular velocity point.
7. Add a new column to the RotationCurve data frame which computes the total error, which is the sum in quadrature of the systematic error, and the error in the measured velocities.

The data frame should have the following aspect:

```
[23]: display.Image("./df.png")
```

[23]:

	radius (kpc)	circ velocity (km/s)	circ velocity error (km/s)	stars circ velocity (km/s)	gas circ velocity (km/s)	syst_error	total_error	DM circ velocity (km/s)
0	0.232710	3.33	1.36	3.048860	2.051611	0.14	1.367187	2.449992
1	0.465421	8.87	4.32	5.447722	4.196565	0.14	4.322268	5.602506
2	0.698131	11.89	4.51	7.264598	6.273282	0.14	4.512172	7.017382
3	0.930841	15.76	5.35	9.047033	6.981577	0.14	5.351831	10.852944
4	1.163550	18.62	5.54	10.911456	6.058803	0.14	5.541769	13.817939

8. Now that you have collected all the data, you can proceed to do the fit. Fit the NFW and the Burkert profiles to the DM circ velocity data. Use the total\_error for that.
9. Compute the goodness of the fit for both the NFW and the Burkert fits.
10. Finally plot your findings

```
[ ]:
```