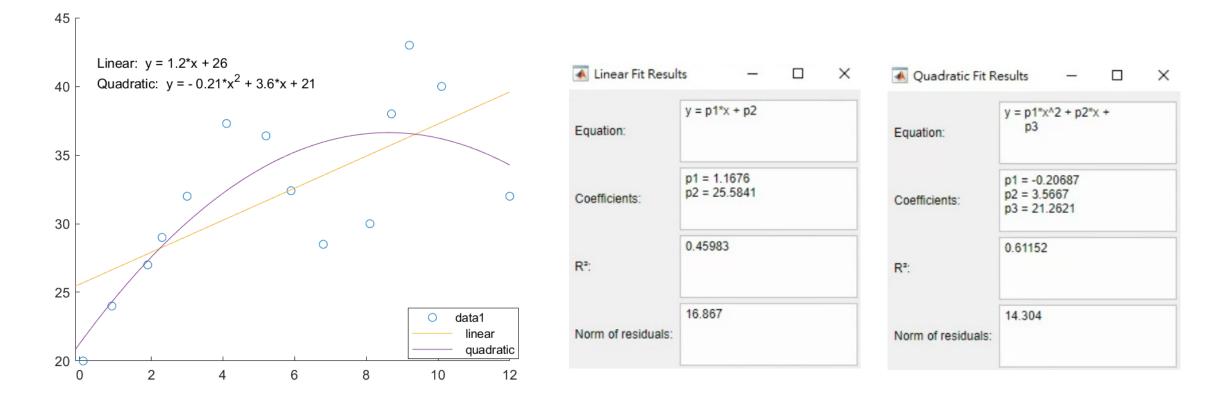
1 Use matlab to find (analytically) the coefficients.

# Result for linear and quadratic regression:



### 2.1

For Conjugate gradient method : cg.py

### Method:

 $\mathbf{r}_0 := \mathbf{b} - \mathbf{A}\mathbf{x}_0$ if  $\mathbf{r}_0$  is sufficiently small, then return  $\mathbf{x}_0$  as the result  $\mathbf{p}_0 := \mathbf{r}_0$  k := 0repeat  $\mathbf{r}_k^{\mathsf{T}} \mathbf{r}_k$ 

$$lpha_k := rac{\mathbf{r}_k^\mathsf{T} \mathbf{r}_k}{\mathbf{p}_k^\mathsf{T} \mathbf{A} \mathbf{p}_k}$$

$$\mathbf{x}_{k+1} := \mathbf{x}_k + \alpha_k \mathbf{p}_k$$

$$\mathbf{r}_{k+1} := \mathbf{r}_k - \alpha_k \mathbf{A} \mathbf{p}_k$$

if  $\mathbf{r}_{k+1}$  is sufficiently small, then exit loop

$$eta_k := rac{\mathbf{r}_{k+1}^\mathsf{T} \mathbf{r}_{k+1}}{\mathbf{r}_k^\mathsf{T} \mathbf{r}_k}$$

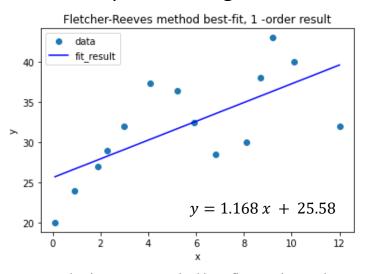
$$\mathbf{p}_{k+1} := \mathbf{r}_{k+1} + \beta_k \mathbf{p}_k$$

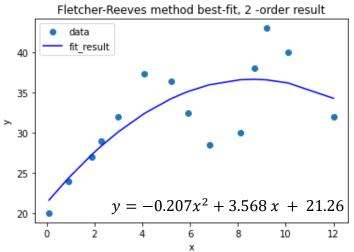
$$k := k + 1$$

end repeat

return  $\mathbf{x}_{k+1}$  as the result

### Result for linear and quadratic regression:





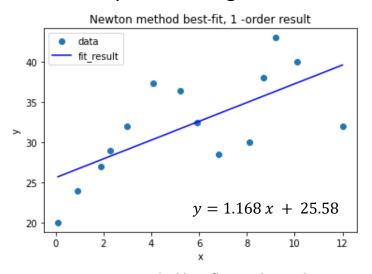
2.2

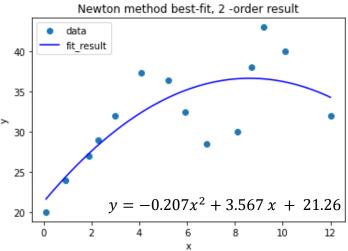
For Newton's method: nt.py

Method:

$$x^{(i+1)} = x^{(i)} - \frac{g(x^{(i+1)})}{g'(x^{(i)})}$$

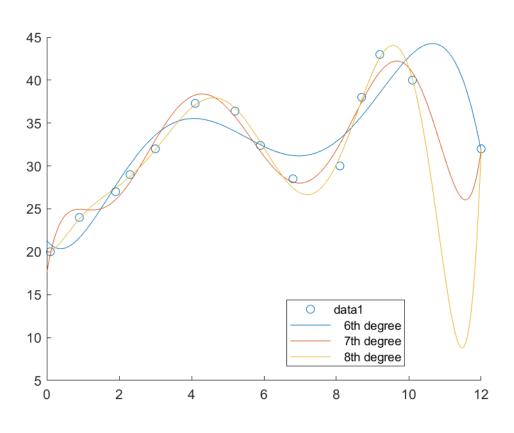
## Result for linear and quadratic regression:





3 Use matlab to find (analytically) the coefficients.

### Result:



Degree = 6  $y = 0.00015846x^6 - 0.013571x^5 + 0.2952x^4 - 2.5211x^3 + 8.4339x^2 - 5.4286x + 21.3294$ 

Degree = 7  $y = 0.00094533x^7 - 0.038215x^6 + 0.59816x^5 - 4.5504x^4 + 17.3089x^3 - 30.8375x^2 + 24.9024x + 17.5479$ 

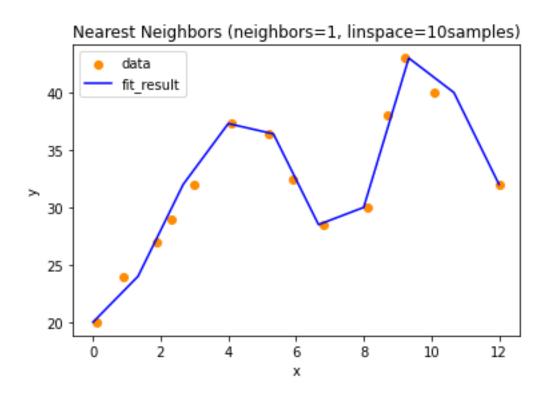
Degree = 8  $y = 0.00022417x^8 - 0.0092746x^7 + 0.15099x^6 - 1.2287x^5 + 5.2657x^4 - 11.6072x^3 + 11.7821x^2 - 0.04113x + 19.9211$ 

當最高項次到6時,函數能剛好隔開每個點。 當最高項次到8時,函數能剛好通過每個點,但是線條有 點overfitting。

如果把資料分成幾類,會比較容易看出有沒有overfitting。

4
Use Nearest Neighbors regression to fit the data: nn.py

### Result:



經過前一題的分析,在避免overfitting的前提下,適度的提高多項式次數,可以找到一個best fit方程式。

但是,如果增加或減少資料(點座標),同一個best fit方程式就可能失效。

為了解決方程式會失效的問題,這裡使用Nearest Neighbors regression model。

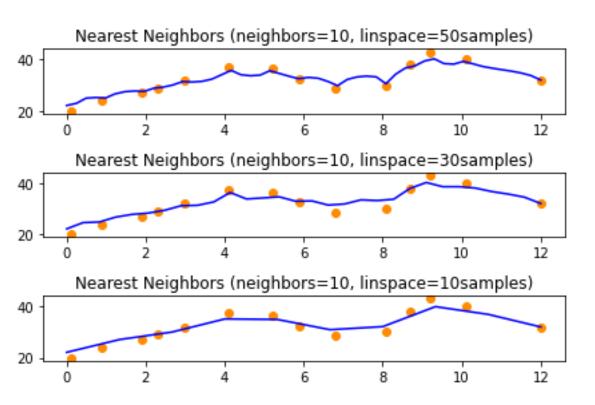
透過持續搜索鄰近同類的方式,可以找到一條很好的best fit line,不因資料增加或減少而失效。

4

Best fit: 14點皆為同一類,應該連在一起,而且不能overfitting

分析:減少鄰居及line space樣本數量,以達到best fit

左下圖 固定鄰居數量(=10),減少line space的樣本數,曲線趨於 平滑(降低overfitting)



右下圖 固定line space樣本數(=10),減少鄰居數量,曲線趨於 精準(14個點都是同一類)

