

Assignment 1

Problem 1

1. **Wrong.** The better a function fits the data, the more it fits the noise, called overfitting. Thus, overfitting happens when the model is too flexible to fit both the significant data and the useless noise. An overfitting model may perform quite well on the training set but poorly on the test set. Zero training error is not necessary.
2. **Decrease.** In the KNN algorithm, when K is small, the model is quite sensitive to changes in the local structure of the dataset, leading to significant variations in performance across different data, which means higher variance. As K increases, the model becomes smoother, the bias in predictions increases, but the variance decreases.
3. $\int (1 - \max q_j(x)) p(x) dx$
4. **The sum of residuals should be zero** in multiple linear regression. In linear regression, to minimize the Residual Sum of Squares, the first derivative is set to zero, resulting in the sum of residuals being zero.
5. **R square can be used to compare two models with same size.** R square means we always choose the model with smaller SSerror which can maximize the R square. However, the complex one with more parameters always has larger R square due to its smaller bias if there are two models with different size. For two same size models, R square works well.
6. **Linear Regression**
Pros: performs well with less data and easy to inference with relatively high interpretability.
Cons: its form could be totally wrong; less flexibility.

KNN Regression

Pros: fits wider range of shapes of function as a non-parametric method.

Cons: it needs a lot of data to work well compared with Linear Regression.

It would be hard to do inference work due to its complicate form.

Problem 2

1. Model summary:

```
In [2]: runcell('Q1', 'C:/Users/ASUS/OneDrive - HKUST Connect/hm/ml/Problem2.py')
OLS Regression Results
=====
Dep. Variable:      Life expectancy      R-squared:                0.839
Model:              OLS                  Adj. R-squared:           0.837
Method:             Least Squares        F-statistic:              422.9
Date:              Fri, 20 Sep 2024       Prob (F-statistic):       0.00
Time:              01:49:12              Log-Likelihood:          -4421.2
No. Observations:   1649                 AIC:                     8884.
Df Residuals:       1628                 BIC:                     8998.
Df Model:           20
Covariance Type:    nonrobust
=====
                    coef      std err          t      P>|t|      [0.025      0.975]
-----
const                308.1207    46.223      6.666    0.000    217.457    398.784
Year                -0.1272     0.023     -5.510    0.000    -0.172    -0.082
Status               0.8865     0.335      2.644    0.008     0.229     1.544
Adult Mortality     -0.0162     0.001    -17.171    0.000    -0.018    -0.014
infant deaths       0.0887     0.011      8.376    0.000     0.068     0.110
Alcohol             -0.1313     0.034     -3.901    0.000    -0.197    -0.065
percentage expenditure  0.0003     0.000      1.691    0.091    -4.83e-05    0.001
Hepatitis B        -0.0033     0.004     -0.732    0.464    -0.012     0.005
Measles            -1.033e-05    1.07e-05    -0.966    0.334    -3.13e-05    1.07e-05
BMI                 0.0318     0.006      5.345    0.000     0.020     0.044
under-five deaths  -0.0666     0.008     -8.682    0.000    -0.082    -0.052
Polio               0.0058     0.005      1.132    0.258    -0.004     0.016
Total expenditure   0.0922     0.040      2.281    0.023     0.013     0.171
Diphtheria         0.0140     0.006      2.387    0.017     0.002     0.026
HIV/AIDS           -0.4481     0.018    -25.174    0.000    -0.483    -0.413
GDP                 2.451e-05    2.83e-05     0.867    0.386    -3.09e-05    7.99e-05
Population          -6.085e-10    1.73e-09    -0.351    0.726    -4.01e-09    2.79e-09
  thinness 1-19 years -0.0058     0.053    -0.111    0.912    -0.109     0.097
  thinness 5-9 years -0.0501     0.052    -0.966    0.334    -0.152     0.052
Income composition of resources 10.4497     0.833     12.549    0.000     8.816    12.083
Schooling           0.8949     0.059     15.142    0.000     0.779     1.011
=====
Omnibus:            31.845    Durbin-Watson:           0.707
Prob(Omnibus):      0.000    Jarque-Bera (JB):        58.052
Skew:               -0.107    Prob(JB):                2.48e-13
Kurtosis:           3.894    Cond. No.:               3.80e+10
=====
```

From this report, we can roughly think *Year, Adult Mortality, infant death, Alcohol, BMI, under-five deaths, HIV/AIDS, Income composition of resources and Schooling* are significant variables affecting the life expectancy. However, there may be some strong multicollinearity problems because we choose too many covariates without further selection, and it is hard to scale the range of all these variables.

2. 95% confidence interval for the coefficients of “Adult Mortality”: - 0.0406 ~ - 0.0352

95% confidence interval for the coefficients of “HIV/AIDS”: - 0.4862 ~ - 0.3734

Although the coefficients are significant negative (95% confidence intervals are lower than zero), we can not be sure that these predictors really have negative impact on the life expectancy. It is because we ignore the impact of other variables when we construct the two factors simple model. For example, some of these ignored variables may have complex interaction effect with the two factors and lead to the final impact on the life expectancy.

3. The 97% confidence interval for the coefficient of “Schooling”: 2.288345 ~ 2.580960

The 97% confidence interval for the coefficient of “Alcohol”: -0.264407 ~ -0.061395

The interval of “Schooling” is above zero, which means that “Schooling” has a positive

impact on the life expectancy with 97% confidence. The interval of “Alcohol” is slightly lower than zero, which means that “Alcohol” has a negative impact to some degree on the life expectancy with 97% confidence.

4. The top seven most influential predictors are HIV/AIDS, Adult Mortality, Schooling, Income composition of resources, under-five deaths, infant deaths, and Year. The summary is showed as below.

```
In [5]: runcell('Q4', 'C:/Users/ASUS/OneDrive - HKUST Connect/hm/ml/Problem2.py')
OLS Regression Results
=====
Dep. Variable:      Life expectancy      R-squared:          0.824
Model:              OLS                  Adj. R-squared:     0.823
Method:              Least Squares        F-statistic:        1096.
Date:                Fri, 20 Sep 2024      Prob (F-statistic):  0.00
Time:                03:02:54             Log-Likelihood:     -4493.3
No. Observations:    1649                 AIC:                9003.
Df Residuals:        1641                 BIC:                9046.
Df Model:             7
Covariance Type:     nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
const	285.5374	45.502	6.275	0.000	196.289	374.786
HIV/AIDS	-0.4498	0.018	-24.615	0.000	-0.486	-0.414
Adult Mortality	-0.0181	0.001	-18.845	0.000	-0.020	-0.016
Schooling	1.0631	0.054	19.677	0.000	0.957	1.169
Income composition of resources	11.9119	0.825	14.440	0.000	10.294	13.530
under-five deaths	-0.0675	0.007	-9.171	0.000	-0.082	-0.053
infant deaths	0.0871	0.010	8.796	0.000	0.068	0.106
Year	-0.1158	0.023	-5.105	0.000	-0.160	-0.071

```
=====
Omnibus:            35.229      Durbin-Watson:        0.657
Prob(Omnibus):      0.000      Jarque-Bera (JB):     70.102
Skew:               -0.081      Prob(JB):             5.99e-16
Kurtosis:           3.997      Cond. No.             1.01e+06
=====
```

5. Predict the Life Expectancy as around 88.38

The 99% Confidence Interval: 85.99 ~ 90.77

6. Full Model AIC: 8884.483951849954

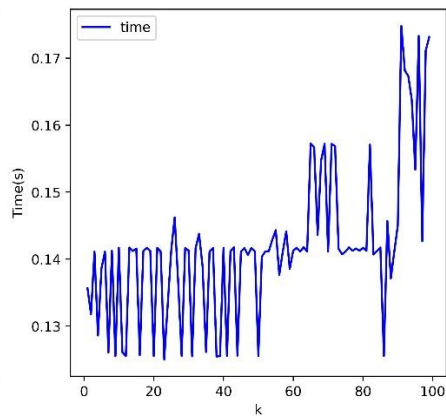
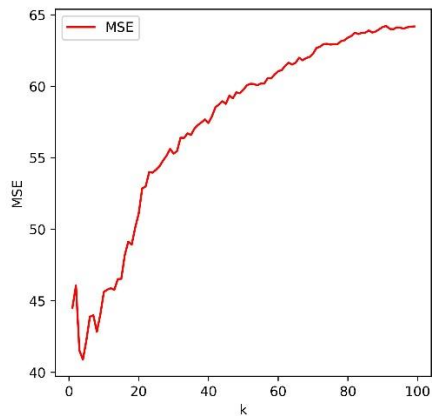
Small Model AIC: 9002.562179694152

The model with smallest AIC is preferred. Thus, it seems that Full Model is preferred than the smaller model.

Problem 3

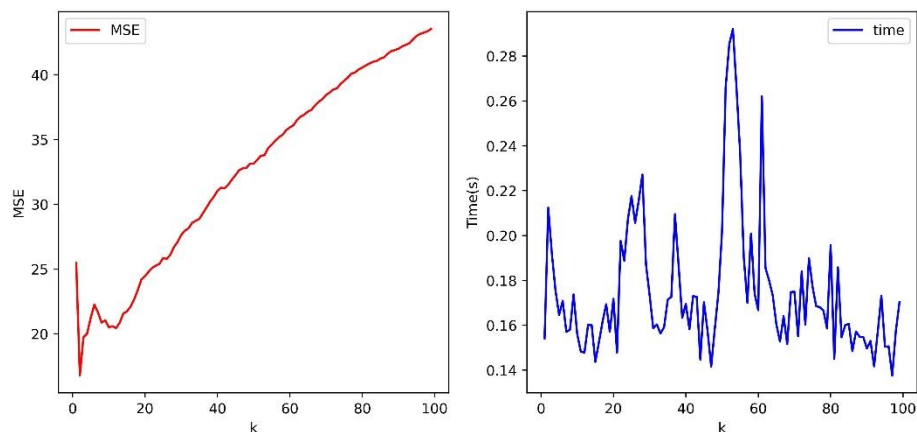
1. Test MSE reach the minimum at around 40.88 when $K=4$. The best value of K is 4.

K (see 'k_range' in code)	MSE (see 'MeanSE')	Time
1	44.51708661417323	0.1268603801727295
2	46.05590551181103	0.1264653205871582
3	41.52323709536307	0.14185070991516113
4	40.88791830708661	0.1414954662322998
5	42.24112440944882	0.1416645050048828
6	43.889407261592304	0.12514352798461914
7	43.985068295034544	0.1412184238433838
8	42.83032357283464	0.12839269638061523
9	44.04344026441139	0.13933968544006348
... (k range from 1 to 100)



- After standardize the predictor variables (excluding the response variable) in both training and test data, test MSE reach the minimum at around 16.77 when K=2. The best value of K is 2.

K (see 'k_range' in code)	MSE (see 'MeanSE')	Time
1	25.46929133858267	0.1542191505432129
2	16.777578740157477	0.2123730182647705
3	19.73487314085739	0.19198060035705566
4	20.019940944881885	0.17559599876403809
5	21.226831496062992	0.16461682319641113
6	22.255448381452318	0.17090654373168945
7	21.654661738711233	0.15705180168151855
8	20.868348917322834	0.15802001953125
9	21.036780402449686	0.17373251914978027
... (k range from 1 to 100)



- Standardization significantly reduced the test MSE, improving the predictive performance of KNN regression, but there was no noticeable change in the model's runtime.

Note: copilot tool is used to Copilot is used to query the necessary third-party library interfaces and some functional functions (such as the `model.pvalues.sort_values()` used in Problem 2).