

Assignment 1
MAT3110 - Introduction to numerical analysis

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1 Exercise 1

Use the QR factorization of A and apply back substitution to R1 to find x. You will need to write your own routine for back substitution but you can use the matlab function $[Q,R] = \text{qr}(A)$; to find Q and R.

Ans:

1.1 Back-substitution implementation

I used the built in function for Q and R calculations from numpy. I implemented the following function for back substitution:

```
def back_subst(A: np.ndarray, b: np.ndarray):
    n = b.shape[0]
    if A.shape[0] != A.shape[1]:
        raise ValueError("Input must be square, douche...")
    x = np.zeros(n)
    x[n-1] = b[n-1] / A[n-1, n-1]

    for i in range(n-2, -1, -1):
        x[i] = (b[i] - A[i, i+1:] @ x[i+1:]) / A[i,i]
    return x
```

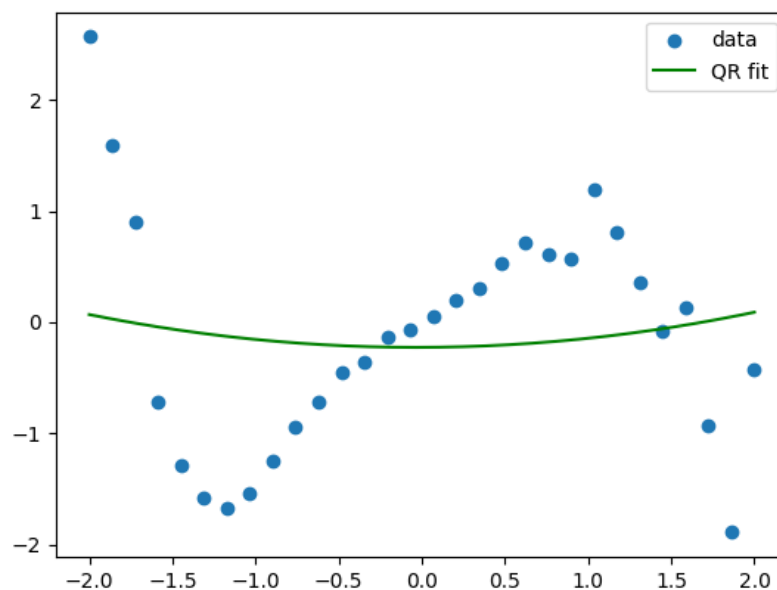
- The code takes input square a upper triangle matrix, and an arbitrary vector b that takes the same shape as the rows and columns of A
- Then I loop backwards from the lower diagonal and up. I have equations of the following form

$$\sum_{j=n}^{n-j} \sum_{i=j+1}^n a_{ji}x_j = b_j$$

And we can solve this by taking the dot product of th

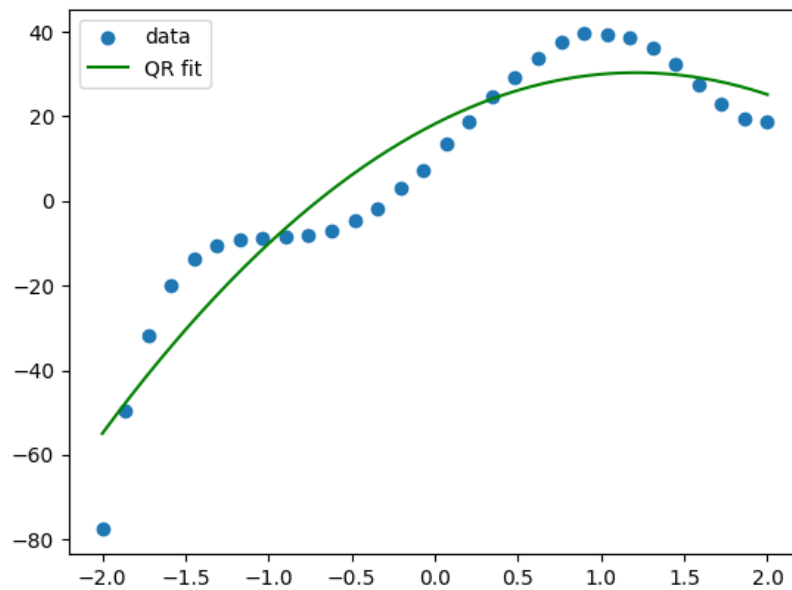
1.2 QR plot for dataset 1 m=3

This produced the following plot for data set 1:



1.3 QR plot for dataset 2 $m=3$

And the following plot for dataset 2:



2 Exercise 2

The $m \times m$ matrix $B = A^T A$ is symmetric and positive definite. Solve the normal equations using the Cholesky factorization RR^T of B . To do this you also need to implement forward substitution and the Cholesky algorithm explained in Lecture 3.

Ans:

2.1 Cholesky implementation

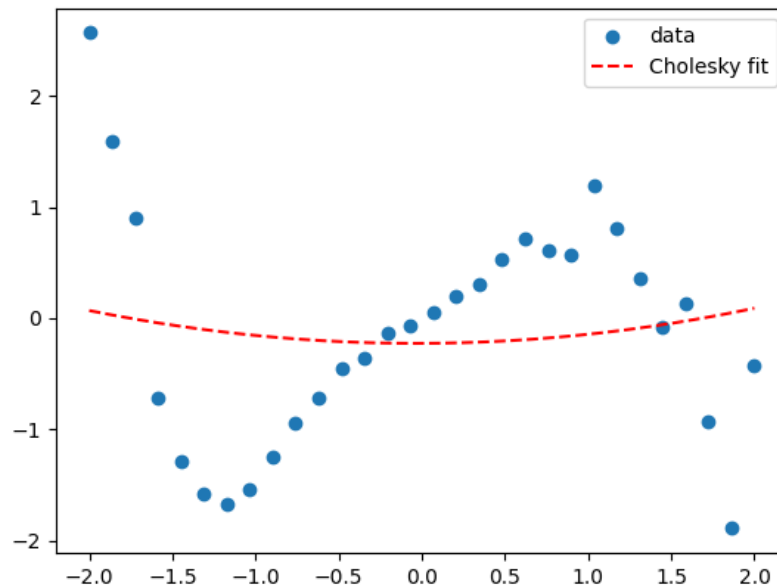
Here is my Cholesky implementation.

```
def cholesky(A):  
    A = A.copy().astype(float)  
    n = A.shape[0]  
    L = np.zeros((n,n))  
    D = np.zeros((n,n))  
    for i in range(n):  
        lk = A[:,i] / A[i,i]  
        L[:,i] = lk  
        D[i,i] = A[i,i]  
        A = A - D[i,i] * np.outer(lk, lk)  
    return L, D
```

Here is the figure for the Cholesky interpolation for dataset 1

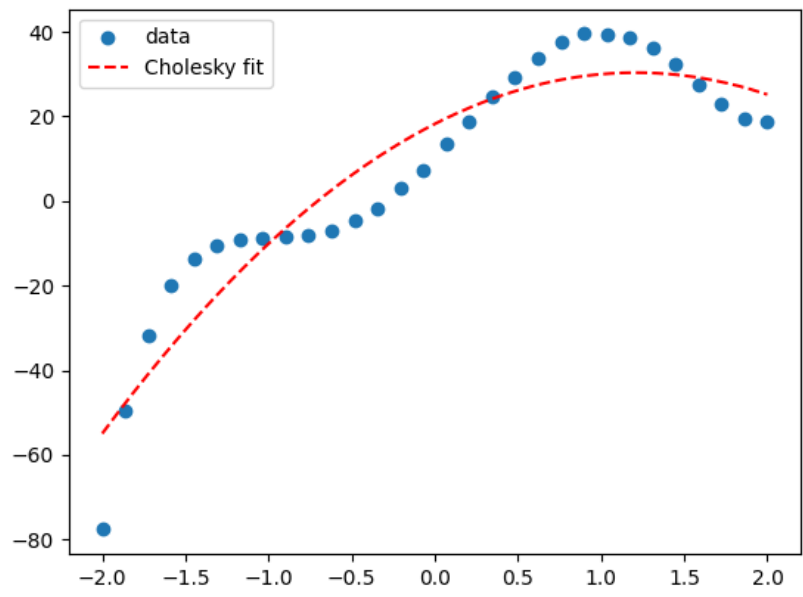
2.2 Cholesky plot for dataset 1 m=3

plott 1 for cholesky



2.3 Cholesky plot for dataset 2 m=3

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3 Exercise 3

Here I will discuss the differences