# Final Term Project

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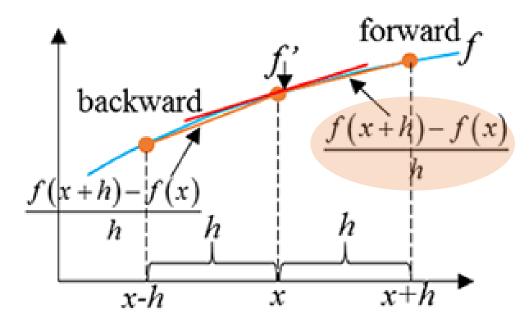
## Research Purpose

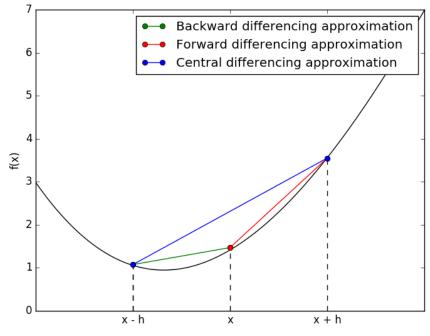
- 'Stansby' used the magnetic field data of Helios E2 and E3 to create a data called 'Corefit' (E2's data = E3's data).
- However, E2 collected data 40 seconds on average, and E3 is 6 seconds.
- Stansby used azimuthal and elevation, but I considered the size of the three component (SSE coordinate).
- Also, I considered the approximation using the finite difference method (chapter 1) learned in class.
- I used Pearson correlation to determine how much correlation it had.

- I used the finite difference matrix I learned in class (chapter 1).
- However, although the book assumed a periodic function, the actual data are not a periodic function, so it is not possible to apply a 'Central finite difference' from the boundary value.

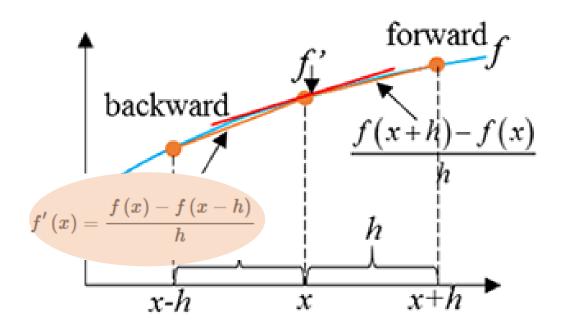
$$h^{-1} \begin{pmatrix} 0 & \frac{1}{2} & & & -\frac{1}{2} \\ -\frac{1}{2} & 0 & \ddots & & \\ & & \ddots & & \\ & & \ddots & 0 & \frac{1}{2} \\ \frac{1}{2} & & -\frac{1}{2} & 0 \end{pmatrix}$$

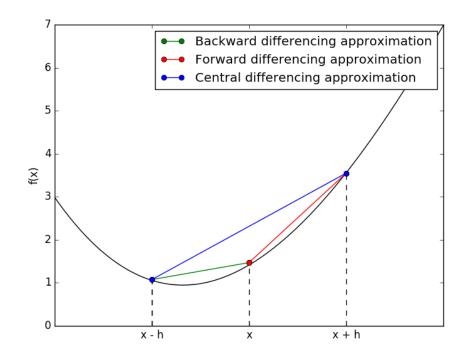
- So I found out through Googleing that there is a method of 'forward' and 'backward'.
- The forward finite difference method is a approximating values ahead of the reference point.





- Conversely, the backward difference method is an approximation of the points located behind the reference point.
- Also, I followed the accuracy order by referring to Wikipedia.





Derivative	Accuracy	-5	-4	-3	-2	-1	0	1	2	3	4	5
	2					-1/2	0	1/2				
	4				1/12	-2/3	0	2/3	-1/12			
1	6			-1/60	3/20	-3/4	0	3/4	-3/20	1/60		

Central

Derivative	Accuracy	0	1	2	3	4	5	6	7	8
	1	-1	1							
1	2	-3/2	2	-1/2						
	3	-11/6	3	-3/2	1/3					
	4	-25/12	4	-3	4/3	-1/4				
	5	-137/60	5	-5	10/3	-5/4	1/5			

**Forward** 

envativ	e Accuracy	-8	-7	-6	-5	-4	-3	-2	-1	0
	1								-1	1
1	2							1/2	-2	3/2
	3						-1/3	3/2	-3	11/6
1	2							-	H	

Backward

#### Code

```
def centered_finite_diff(N,h): #centered
 line = np.zeros(N)
 line[1] = 2/3; line[N-1] = -2/3; line[2] = -1/12; line[N-2] = 1/12
 matrix = 1/h * toep(-line, line)
 return matrix
def forward_finite_diff(N,h): #forward
 line = np.zeros(N)
 Iine[0] = -25/12; Iine[1] = 4; Iine[2] = -3; Iine[3] = 4/3; Iine[4] = -1/4
 line2 = np.zeros(N)
 line2[0] = -25/12; line2[N-1] = 4; line2[N-2] = -3; line2[N-3] = 4/3; line2[N-4] = -1/4
 matrix = 1/h * toep(line2, line)
 return matrix
def backward_finite_diff(N,h): #backward
 line = np.zeros(N)
 line[0] = 25/12; line[1] = -4; line[2] = 3; line[3] = -4/3; line[4] = 1/4
 line2 = np.zeros(N)
 line2[0] = 25/12 ; line2[N-1] = -4 ; line2[N-2] = 3 ; line2[N-3] = -4/3 ; line2[N-3]
 matrix = 1/h * toep(line,line2)
 return matrix
```

$$h^{-1} \begin{pmatrix} & & \ddots & & \frac{1}{12} & -\frac{2}{3} \\ & & \ddots & -\frac{1}{12} & & \frac{1}{12} \\ & & \ddots & \frac{2}{3} & \ddots & \\ & & \ddots & 0 & \ddots & \\ & & \ddots & -\frac{2}{3} & \ddots & \\ -\frac{1}{12} & & & \frac{1}{12} & \ddots & \\ & & & \frac{2}{3} & -\frac{1}{12} & & \ddots & \end{pmatrix}$$

Accuracy	0	1	2	3	4	5
1	-1	1				
2	-3/2	2	-1/2			
3	-11/6	3	-3/2	1/3		
4	-25/12	4	<del>-</del> 3	4/3	-1/4	

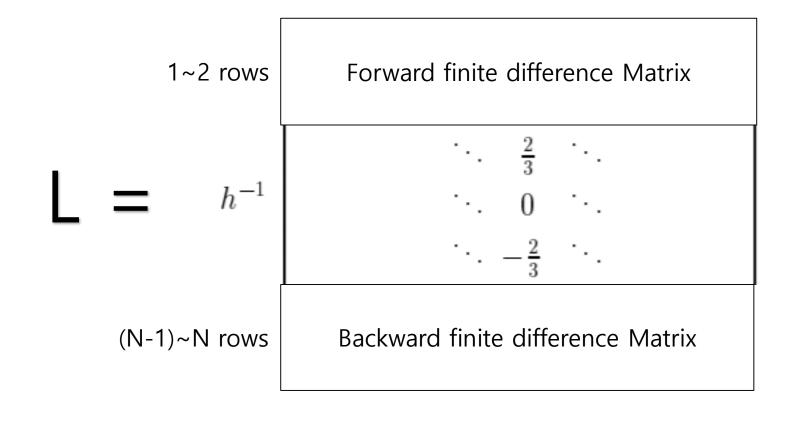
#### Code

```
N = len(E2_x)
h = 1/90

A = centered_finite_diff(N,h)
B = forward_finite_diff(N,h)
C = backward_finite_diff(N,h)

L = np.zeros_like(A)
L[0] = B[0] ; L[1] = B[1]
L[N-1] = C[N-1] ; L[N-2] = C[N-2]
L[2:N-2] = A[2:N-2]
```

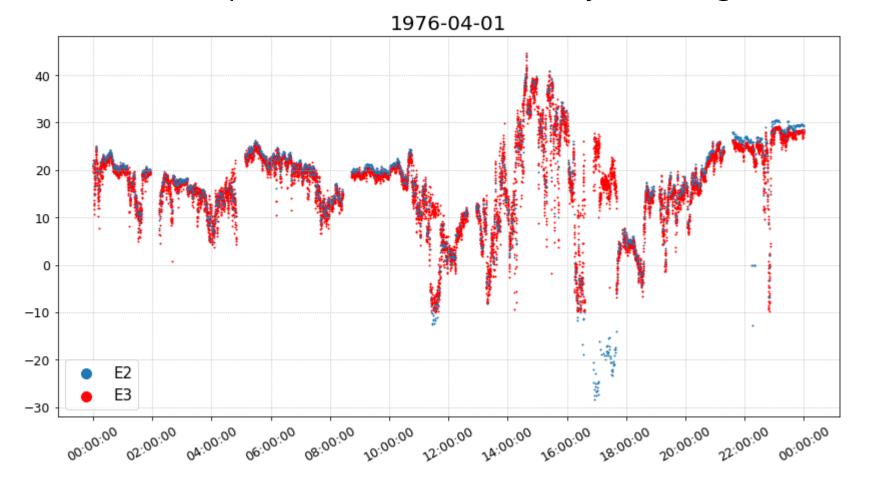
#### All Accuracy = 4



-2/3

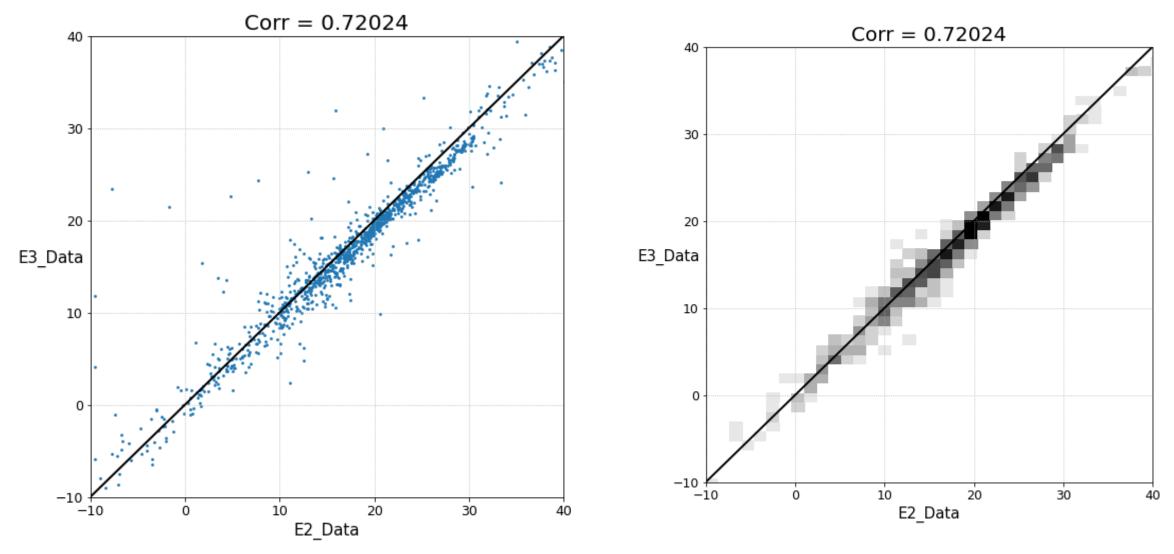
The approximation was calculated using this L matrix.

First, I compared E2 and E3 data by drawing them in a scatter plot.



X-axis in SSE coordinate

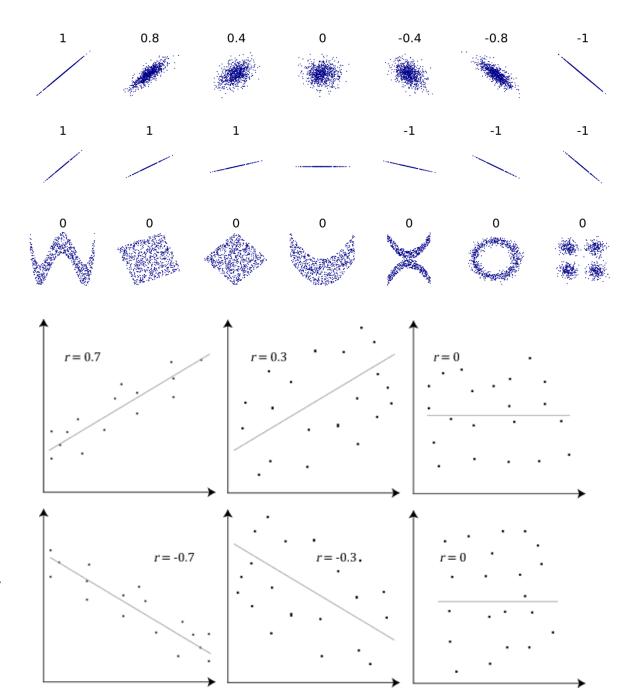
There is a similar tendency.



The black line represents y=x.

#### Pearson C.C

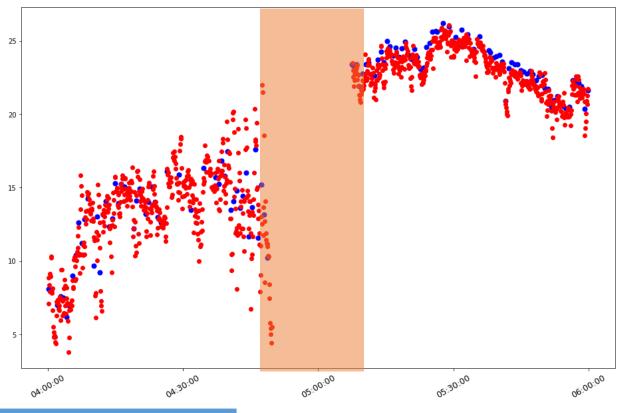
- I understood that x increases as positive correlation y increases.
- Conversely, the negative correlation was understood to decrease y as x increased.
- I haven't studied much about this part because it's statistical. I'm sorry.



Second, I tried to interpolation and regression

Because there are empty gaps and not uniform in the data, interpolation was implemented as a whole.

Later, an approximation was obtained using a regression equation.



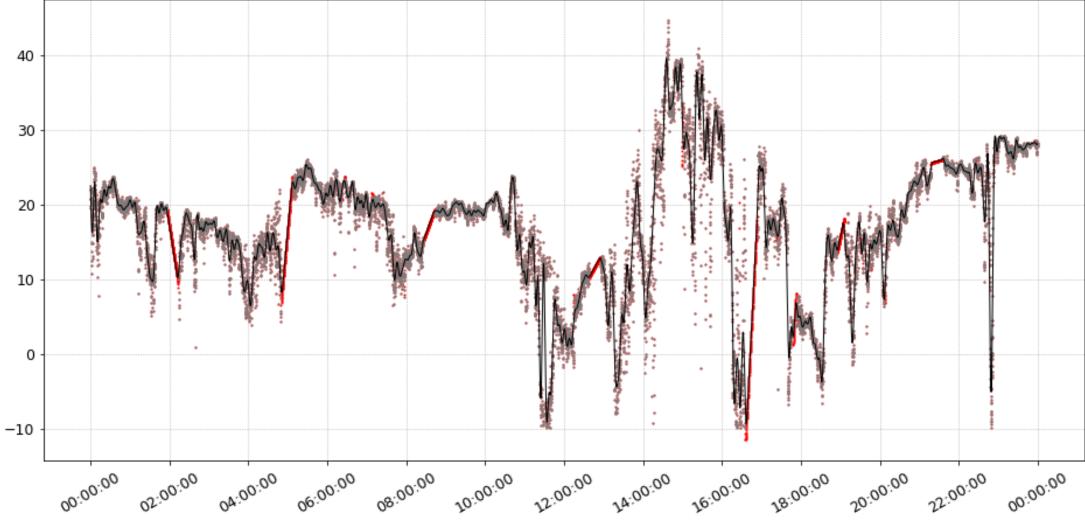
	E2	<b>E</b> 3
Grid Point	Ave 40 sec	Ave 6 sec
No. Strange Data (1976.4.1)	152 / 1772	45 / 13004

Not uniform

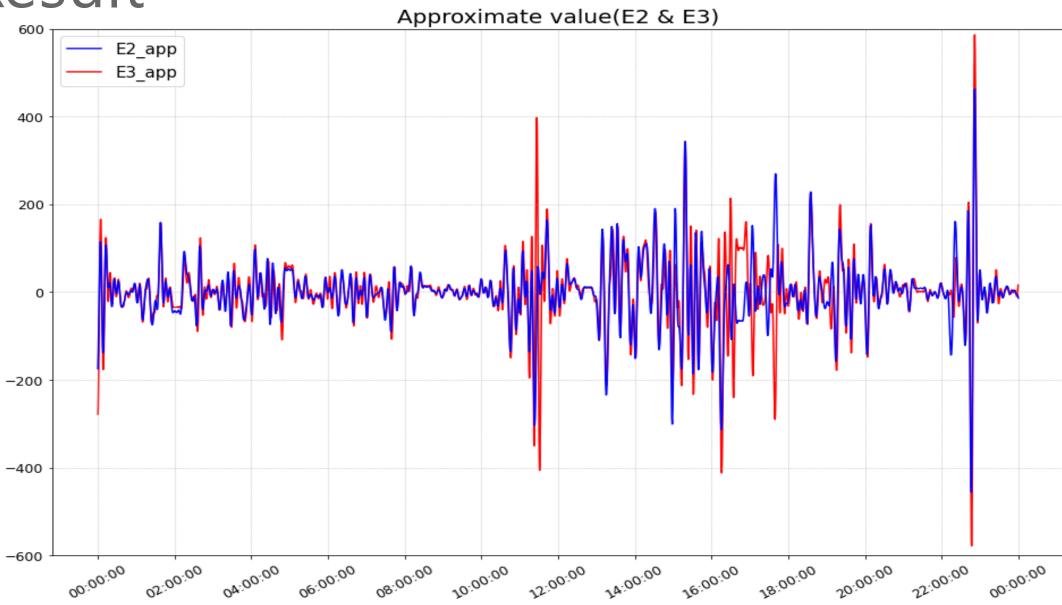
#### Result E2(40sec) 40 30 20 10 0 -10 -20 -30

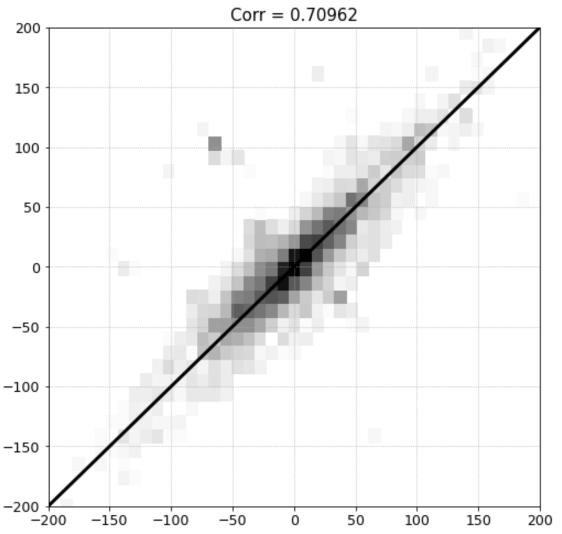
Black line: Reg(SVR), Grey point: Origin point, Blue point: Inter point



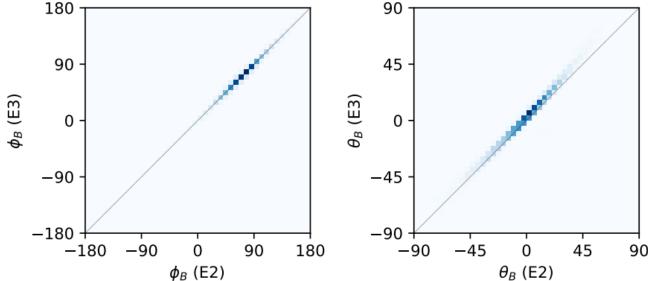


Black line: Reg(SVR), Grey point: Origin point, Red point: Inter point





If the count in bin is less than five, it is not drawn.



It can be seen that it is more widely spread than the values considered for azimuthal and elevation.

#### Discussion

- The time from 16H to 18H tends to not fit well. This is thought to be the difference between grid period or by measurement method.
- The Pearson correlation coefficient is 0.7 to 0.9 and indicates good correlation just by the magnitude of each vector component other than the azimuthal and elevation.
- It seems that data from empty part can be interpolated through machine learning (Requires further study).
- 'Corefit' using E2 and E3 data by Stansby is thought to be reasonable enough data.
- Regardless of the grid period, the data in E2 and E3 have similar values and the approximate differences have similar values, so the two data can be used interchangeably.