Green,
$$\frac{d}{dt}$$
 $\frac{c_1}{c_1} = 0$ $\frac{c_2}{c_2}$ $\frac{c_3}{c_4}$ $\frac{c_4}{c_4}$ $\frac{c_4}$

Given

$$\frac{d}{dt} \begin{bmatrix} e_1 \\ e_1 \\ e_2 \\ e_2 \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & -4\alpha & 4\alpha & -2\alpha(\pi) \\ 0 & 0 & 0 & 1 \\ 0 & -2\alpha(\pi) & 2\alpha(\pi) & -2\alpha(\pi) \\ \hline I_2 V & \overline{I_2} & \overline{I_2} \\ \end{bmatrix}$$

$$+ \begin{bmatrix} 0 \\ 2\alpha \\ \overline{M} \\ 0 \\ 2\alpha \overline{I_2} \\ \end{bmatrix}$$

$$+ \begin{bmatrix} 0 \\ 2\alpha \overline{I_2} \\ \overline{I_2} \\ \end{bmatrix}$$
Here $V = \frac{2}{x}$

$$V = \frac{39}{x} \quad | S = \frac{1}{x} \quad | S =$$

Iz = 25854 M = 1888.6 kg

$$A = \begin{cases} 0 & 1 & 0 & 0 \\ 0 & -\frac{42.359}{V} & 42.359. & -\frac{3.388}{V} \\ 0 & 0 & 0 & 1. \\ 0 & -0.2475 & 0.2475. & -6.7 \\ \hline V & V & 0 & 0 & 0 \\ \hline V & 0 & 0 & 0 & 0 \\ V & 0 & 0 & 0 & 0 \\ \hline V & 0 & 0 & 0 & 0 \\ \hline V & 0 & 0 & 0 & 0 \\ V & 0 & 0 & 0 & 0 \\ \hline V & 0 & 0 & 0 & 0 \\ \hline V & 0 & 0 & 0 & 0 \\ V & 0 & 0 & 0 & 0 \\ \hline V & 0 & 0 & 0 & 0 \\ \hline V & 0 & 0 & 0 & 0 \\ V & 0 & 0 & 0 & 0 \\ \hline V & 0 & 0 & 0 & 0 \\ \hline V & 0 & 0 & 0 & 0 \\ V$$

$$B = \begin{cases} 0 & 0 & 0 & 0 \\ 21.1795 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 2.398 & 0 & 0 & 0 & 0 \end{bmatrix}$$

From the code, we can see for all the Mose velocity the system is both controlled by a Observable.

From the figure, so as the velocity increases Q20 we can see That for pole 1-3, the values start from negative to zero hence becoming less stable & for pole 4 the value increases from 0 to high positive value, which also shows The systèm becomy less stuble. But in the losio (Gy) us velocity plot we can see the log rate approaches zero at high velocity, which shows the system becoming mose & more controllable, as & value increases a pandochillrent

```
In [14]:
```

```
import numpy as np
import matplotlib.pyplot as plt
```

```
In [15]:
```

```
Velo = [2, 5, 8]
```

In [16]:

```
for v in Velo:
    A= np.array([[0, 1], [0, 0]])
    print(A.shape)
    C = np.identity(2)
    B = np.expand_dims(np.array([0, 1/1888.6]).T, axis =1)

P = np.hstack((B, A@B, A@A@B, A@A@A@B))
#print(P.shape)

Q = np.vstack((C, C@A, C@A@A, C@A@A@A))
#print(Q.shape)
    rankP = np.linalg.matrix_rank(P)
    rankQ = np.linalg.matrix_rank(Q)
    print(f'For Velocity = {v} m/s\n')
    print(f'Rank of Controllability matrix = {rankP}')
    print(f'Rank of Observability matrix = {rankQ} \n')
```

```
(2, 2)
For Velocity = 2 m/s

Rank of Controllability matrix = 2
Rank of Observability matrix = 2
(2, 2)
For Velocity = 5 m/s

Rank of Controllability matrix = 2
Rank of Observability matrix = 2
(2, 2)
For Velocity = 8 m/s

Rank of Controllability matrix = 2
Rank of Observability matrix = 2
```

In [17]:

```
for v in Velo:
   A = np.array([[0, 1, 0, 0],
                 [0, -42.359/v, 42.359, -3.388/v],
                 [0, 0, 0, 1],
                 [0, -0.2475/v, 0.2475, -6.7/v]])
   C = np.identity(4)
   B = np.expand_dims(np.array([0, 21.1795, 0, 2.398]).T, axis =1)
   P = np.hstack((B, A@B, A@A@B, A@A@A@B))
   #print(P.shape)
   Q = np.vstack((C, C@A, C@A@A, C@A@A@A))
   #print(Q.shape)
   rankP = np.linalg.matrix_rank(P)
   rankQ = np.linalg.matrix_rank(Q)
    print(f'For Velocity = {v} m/s\n')
    print(f'Rank of Controllability matrix = {rankP}')
    print(f'Rank of Observability matrix = {rankQ} \n')
For Velocity = 2 \text{ m/s}
```

```
Rank of Controllability matrix = 4
Rank of Observability matrix = 4
For Velocity = 5 m/s

Rank of Controllability matrix = 4
Rank of Observability matrix = 4
For Velocity = 8 m/s

Rank of Controllability matrix = 4
Rank of Observability matrix = 4
```

From the results above we can see the system is controllable and observable in all the velocity

Part 2

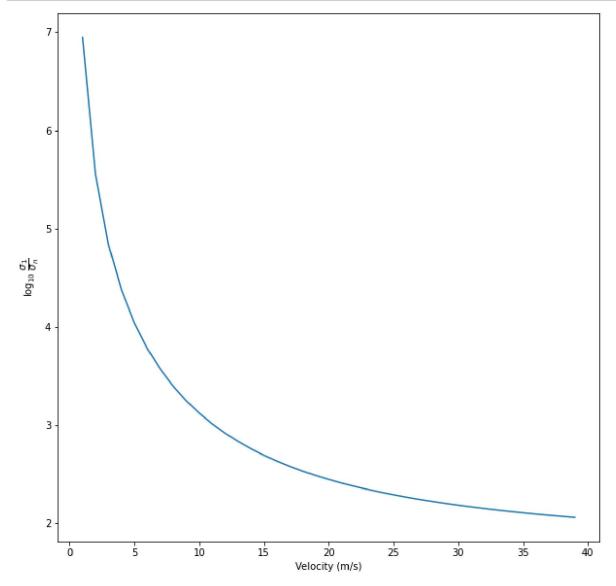
In [18]:

```
pole1 = []
pole2 = []
pole3 = []
pole4 =[]
log =[]
for i in range(1, 40):
    v = i
    A = np.array([[0, 1, 0, 0],
                 [0, -42.359/v, 42.359, -3.388/v],
                 [0, 0, 0, 1],
                 [0, -0.2475/v, 0.2475, -6.7/v]])
    B = np.expand dims(np.array([0, 21.1795, 0, 2.398]).T, axis =1)
    C = np.array([1, 1, 1, 1])
    \#C = np.identity(4)
    D = np.array([0])
    \#D = np.zeros((4,4))
    P = np.hstack((B, A@B, A@A@B, A@A@A@B))
    #print(P.shape)
    Q = np.vstack((C, C@A, C@A@A, C@A@A@A))
    U, sigma, V = np.linalg.svd(P)
    sigma1 = np.max(sigma)
    sigman = np.min(sigma)
    log.append(np.log10(sigma1 / sigman))
    Eig = np.linalg.eig(A)
    #print(Eig[0])
    #print(Eig[1])
    #S = control.StateSpace(A, B, C, D)
    #poles = control.pole(S)
    pole1.append(Eig[0][0].real)
    pole2.append(Eig[0][1].real)
    pole3.append(Eig[0][2].real)
    pole4.append(Eig[0][3].real)
```

Sigma plot

In [19]:

```
plt.figure(figsize = (10, 10))
plt.plot(np.arange(1, 40), log)
plt.xlabel('Velocity (m/s)')
plt.ylabel('$\log_{10}$ $\dfrac{\sigma_1}{\sigma_n}$')
plt.show()
```



Q1 - Jupyter Notebook

```
In [20]:
```

```
fig, ax = plt.subplots(2, 2, figsize=(10,10))

ax[0,0].set_title('pole 1')
ax[0,0].plot(np.arange(1, 40), pole1)

ax[1,0].set_title('pole 2')
ax[1,0].plot(np.arange(1, 40), pole2)

ax[0,1].set_title('pole 3')
ax[0,1].plot(np.arange(1, 40), pole3)

ax[1,1].set_title('pole 4')
ax[1,1].plot(np.arange(1, 40), pole4)
for a in ax.flat:
    a.set(xlabel='Velocity', ylabel='Real part of poles')
```

