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In [2]: import numpy as np
        from scipy.io import loadmat
```

### Question 1-a

$$V = \begin{bmatrix} 1 & X & X \\ X & 2 & 4 \\ -1 & 2 & X \\ X & -2 & X \end{bmatrix}$$

1 rank matrix

$$V = \begin{bmatrix} 1 & -2 & -4 \\ -1 & 2 & 4 \\ -1 & 2 & 4 \\ 1 & -2 & -4 \end{bmatrix}$$

### Question 1-b

We need to know at least 4 elements to fill in the rank 1, so we can miss at most 8 and still complete, so we can miss at minimum 9 and not be able to complete it.

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## Question 2

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In [19]: Xtrue = loadmat("incomplete.mat")["Xtrue"]
        Y1 = loadmat("incomplete.mat")["Y1"]
        Y2 = loadmat("incomplete.mat")["Y2"]
        Y3 = loadmat("incomplete.mat")["Y3"]

        def ItSingValThresh(Y, r=1):
            """
            Iterative Singular Value Thresholding function for Matrix Completion
            """
            tol = 10**(-3) # difference between iterates at termination
            max_its = 100;
            n,p = Y.shape
            X = np.array(Y) #make a copy so operations do not mutate the original
            X[np.isnan(X)] = 0 # Fill in missing entries with zeros

            err = 10**6
            itt = 0
```

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while err > tol and itt < max_its:
    U,s,VT = np.linalg.svd(X, full_matrices=False)
    V, S = VT.T, np.diag(s)
    Xnew = U[:, :r] @ np.diag(s[:r]) @ V.T[:, :] ### Complete this line
    for i in range(n):
        for j in range(p):
            if ~np.isnan(Y[i,j]): #replace Xnew with known entries
                Xnew[i,j] = Y[i,j]
    err = np.linalg.norm(X-Xnew, 'fro')
    X = Xnew
    itt += 1
return X

recovered_Y1 = ItSingValThresh(Y1, r=2)
recovered_Y2 = ItSingValThresh(Y2, r=2)
recovered_Y3 = ItSingValThresh(Y3, r=2)

num_nans_Y1 = np.sum(np.isnan(Y1))
num_nans_Y2 = np.sum(np.isnan(Y2))
num_nans_Y3 = np.sum(np.isnan(Y3))

print(f"Number of NaNs in Y1: {num_nans_Y1}")
print(f"Number of NaNs in Y2: {num_nans_Y2}")
print(f"Number of NaNs in Y3: {num_nans_Y3}")

print("Difference between recovered Y1 and Xtrue:", np.linalg.norm(recovered_Y1 - Xtrue))
print("Difference between recovered Y2 and Xtrue:", np.linalg.norm(recovered_Y2 - Xtrue))
print("Difference between recovered Y3 and Xtrue:", np.linalg.norm(recovered_Y3 - Xtrue))

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Number of NaNs in Y1: 136
Number of NaNs in Y2: 76
Number of NaNs in Y3: 16
Difference between recovered Y1 and Xtrue: 87.24667705099742
Difference between recovered Y2 and Xtrue: 0.00473559952740616
Difference between recovered Y3 and Xtrue: 0.0007153218655176282

```

We can see that less NaN's means less error

## Question 2-b

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In [20]: recovered_Y1_rank3 = ItSingValThresh(Y1, r=3)
recovered_Y2_rank3 = ItSingValThresh(Y2, r=3)
recovered_Y3_rank3 = ItSingValThresh(Y3, r=3)

print("Difference between recovered Y1 and Xtrue (rank=3):", np.linalg.norm(recovered_Y1_rank3 - Xtrue))
print("Difference between recovered Y2 and Xtrue (rank=3):", np.linalg.norm(recovered_Y2_rank3 - Xtrue))
print("Difference between recovered Y3 and Xtrue (rank=3):", np.linalg.norm(recovered_Y3_rank3 - Xtrue))

Difference between recovered Y1 and Xtrue (rank=3): 128.77804846772108
Difference between recovered Y2 and Xtrue (rank=3): 48.97940976510773
Difference between recovered Y3 and Xtrue (rank=3): 20.785069891601783

```

The error is worse when we use the wrong rank

