Is the data missing at random?

DEALING WITH MISSING DATA IN PYTHON



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Possible reasons for missing data

Note - (variable o data field or column in a DataFrame)

- Values simply missing at random instances or intervals in a variable
- Values missing due to another variable
- Values missing due to the missingness of the same or another variable

Types of missingness

- 1. Missing Completely at Random (MCAR)
- 2. Missing at Random (MAR)
- 3. Missing Not at Random (MNAR)



Missing Completely at Random(MCAR)

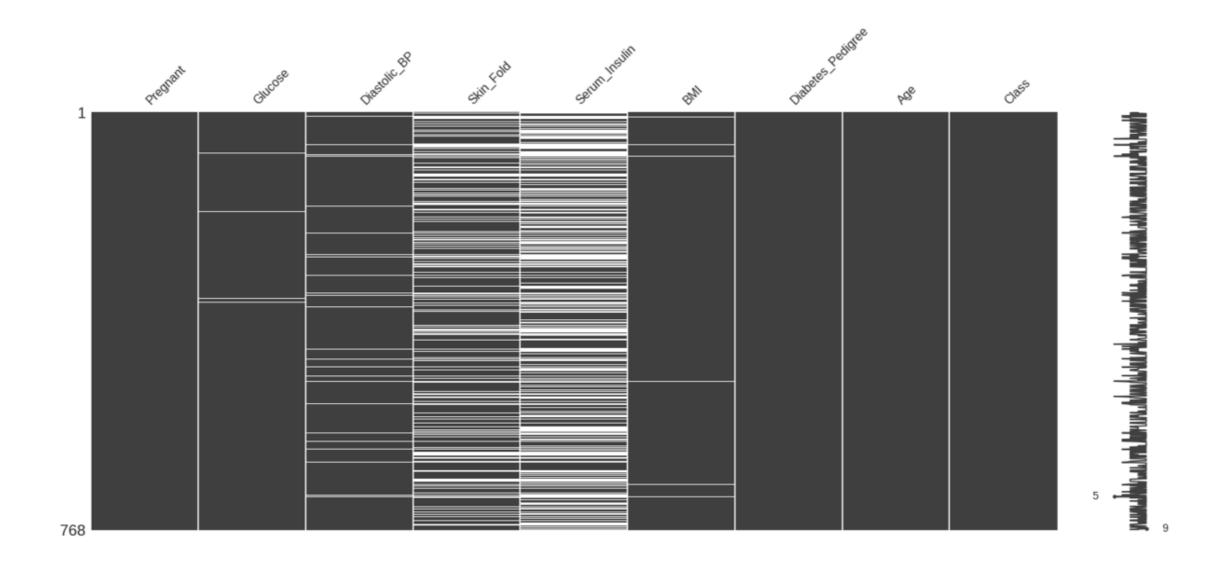
Definition:

"Missingness has no relationship between any values, observed or missing"



MCAR - An example

msno.matrix(diabetes)





Missing at Random(MAR)

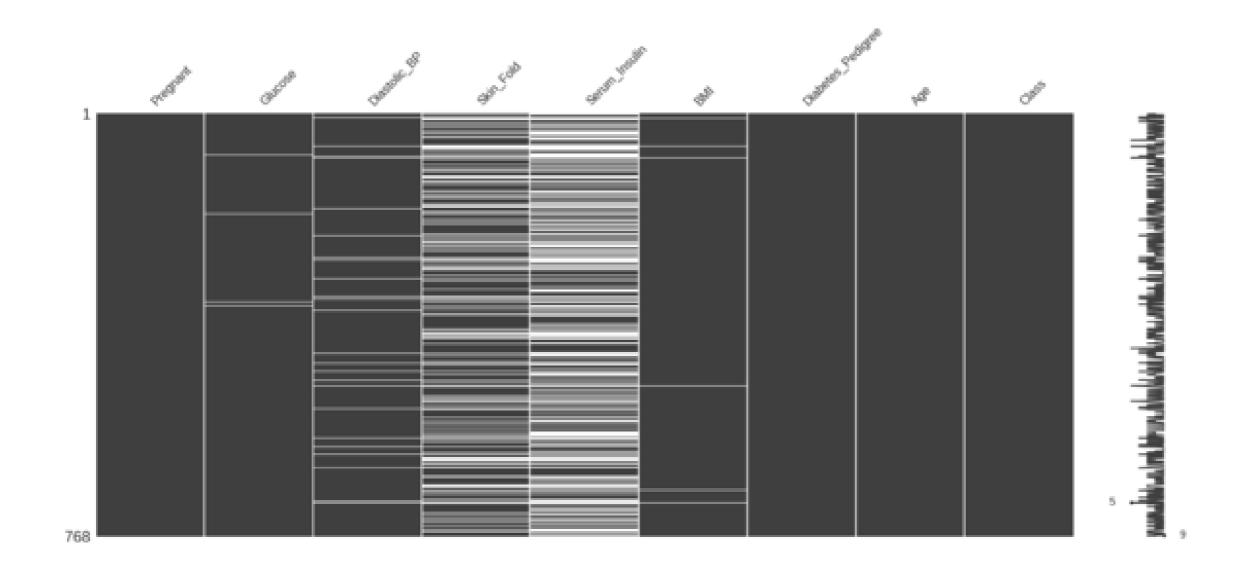
Definition:

"There is a systematic relationship between missingness and other observed data, but not the missing data"



MAR - An example

msno.matrix(diabetes)





Missing not at Random(MNAR)

Definition:

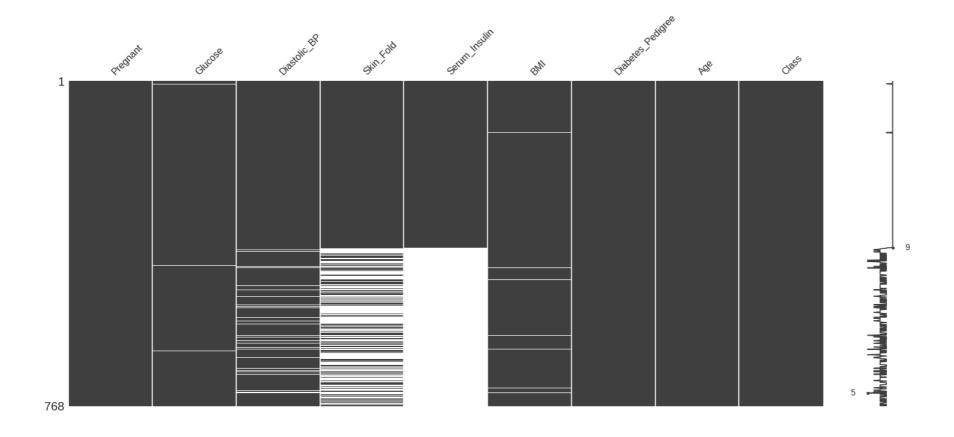
"There is a relationship between missingness and its values, missing or non-missing"



MNAR - An example

Missingness pattern of the diabetes sorted by Serum_Insulin

```
sorted = diabetes.sort_values('Serum_Insulin')
msno.matrix(sorted)
```



Summary

- Possible reasons for missingness
 - Missing Completely at Random (MCAR),
 - Missing at Random (MAR) or
 - Missing Not at Random (MNAR)
- Detecting missingness pattern by sorting the variables
- Mapping missingness to MCAR, MAR & MNAR

Let's practice!

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Finding patterns in missing data

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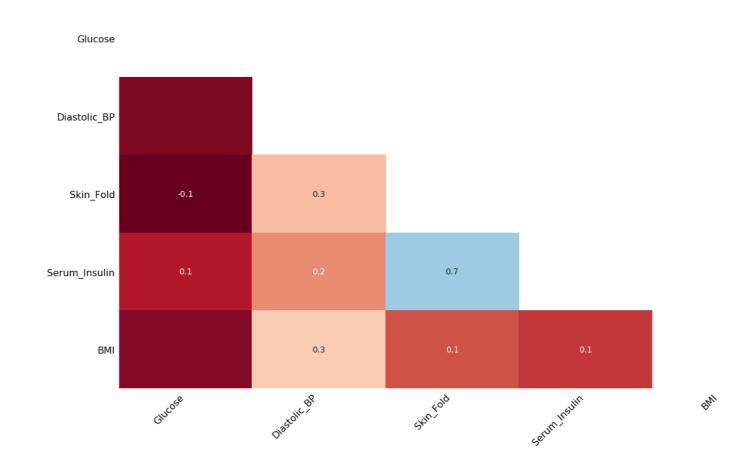
Finding correlations between missingness

- Missingness heatmap or correlation map
- Missingness dendrogram

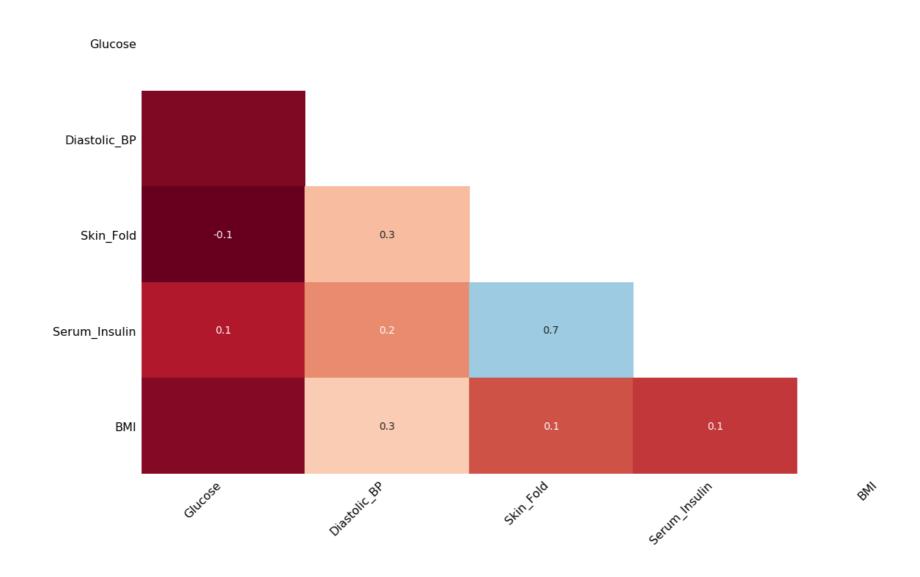


Missingness Heatmap

- Graph of correlation of missing values between columns
- Explains the dependencies of missingness between columns



```
import missingno as msno
diabetes = pd.read_csv('pima-indians-diabetes data.csv')
msno.heatmap(diabetes)
```

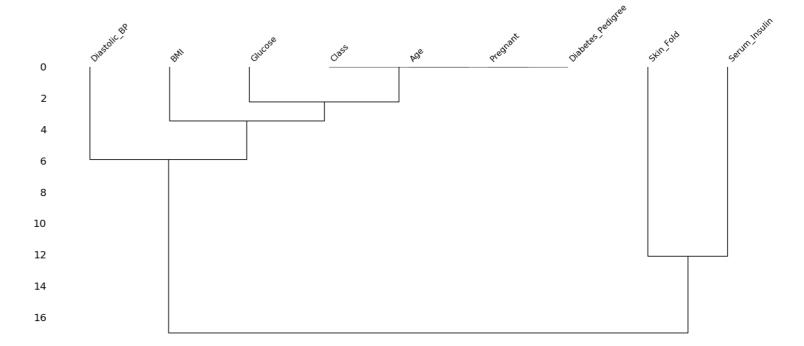


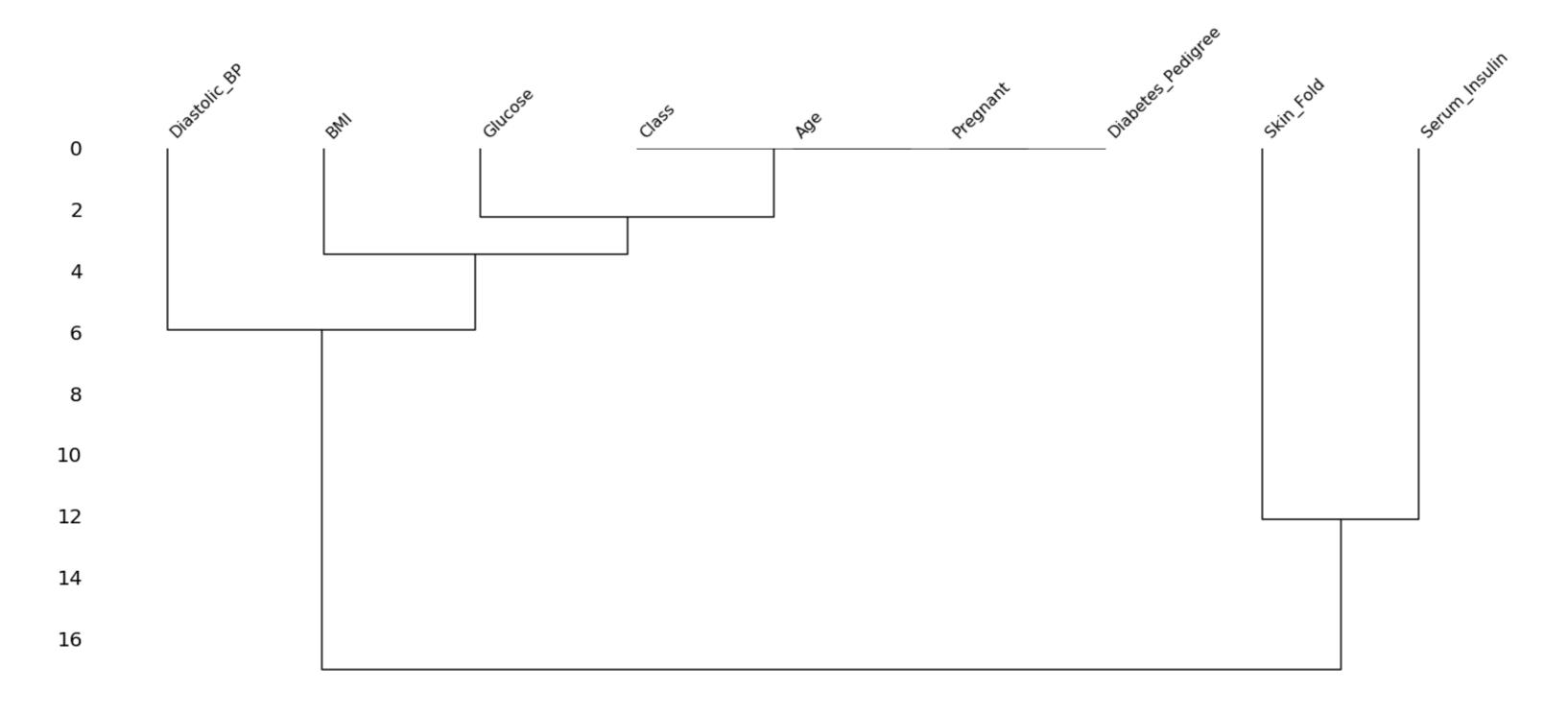


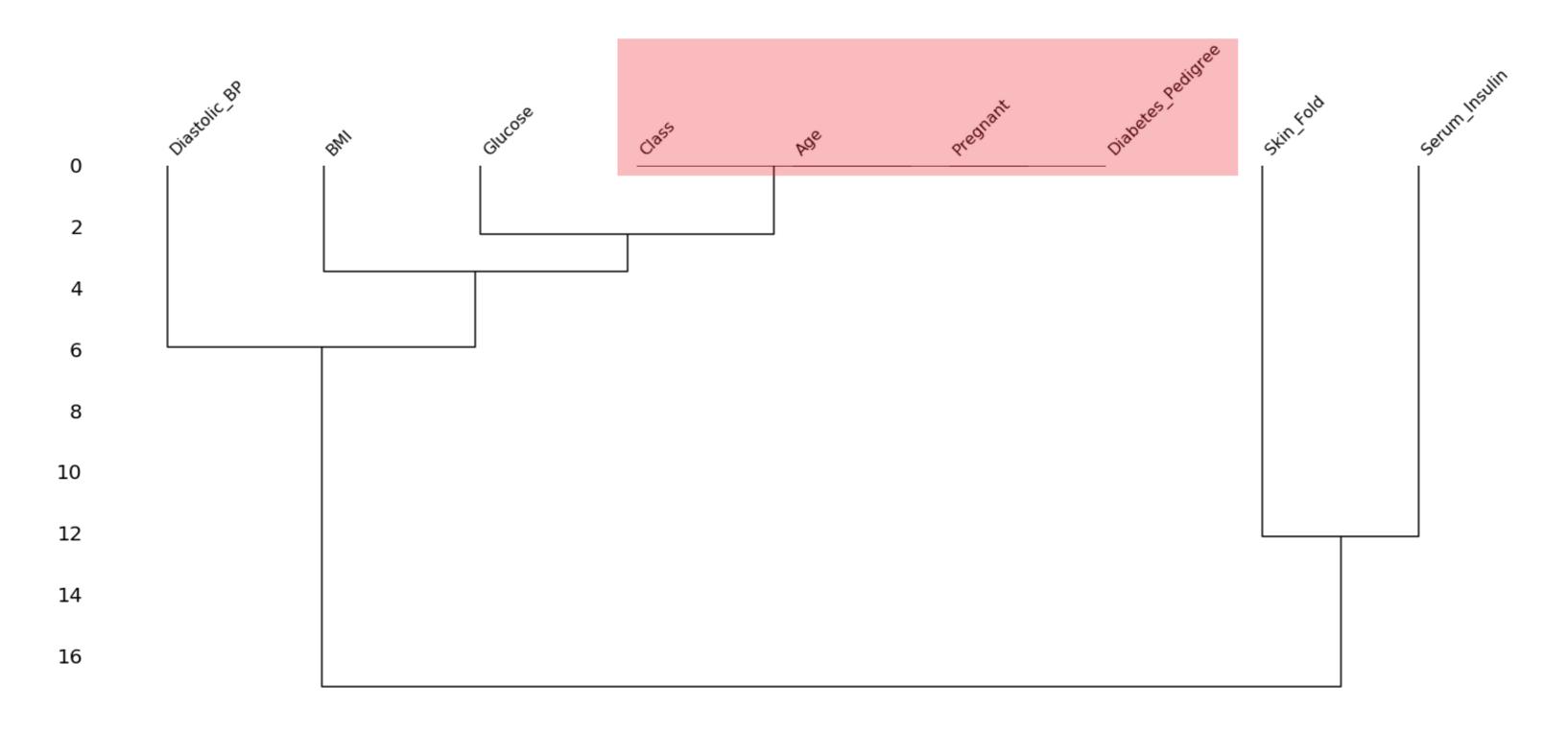
Missingness Dendrogram

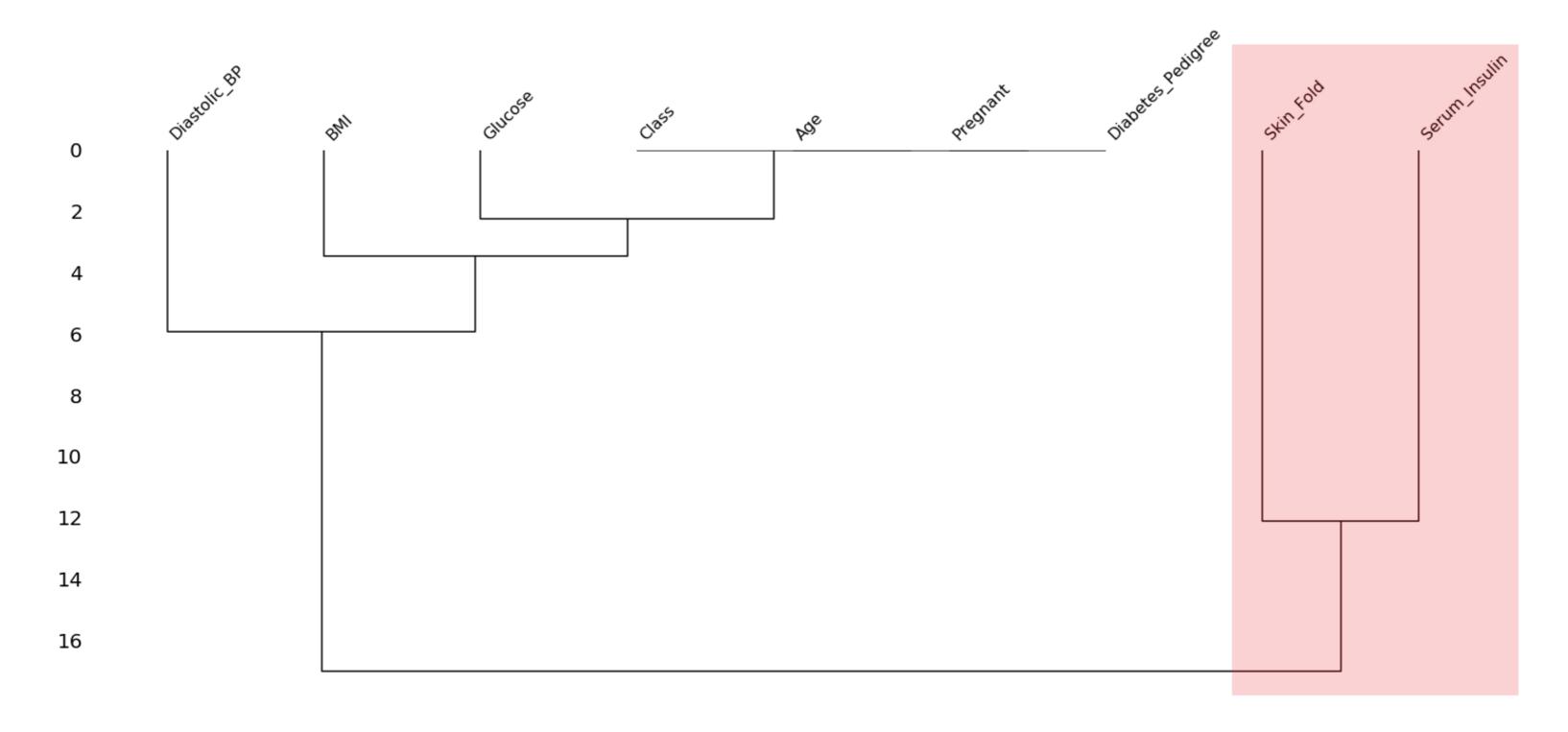
- Tree diagram of missingness
- Describes correlation of variables by grouping them

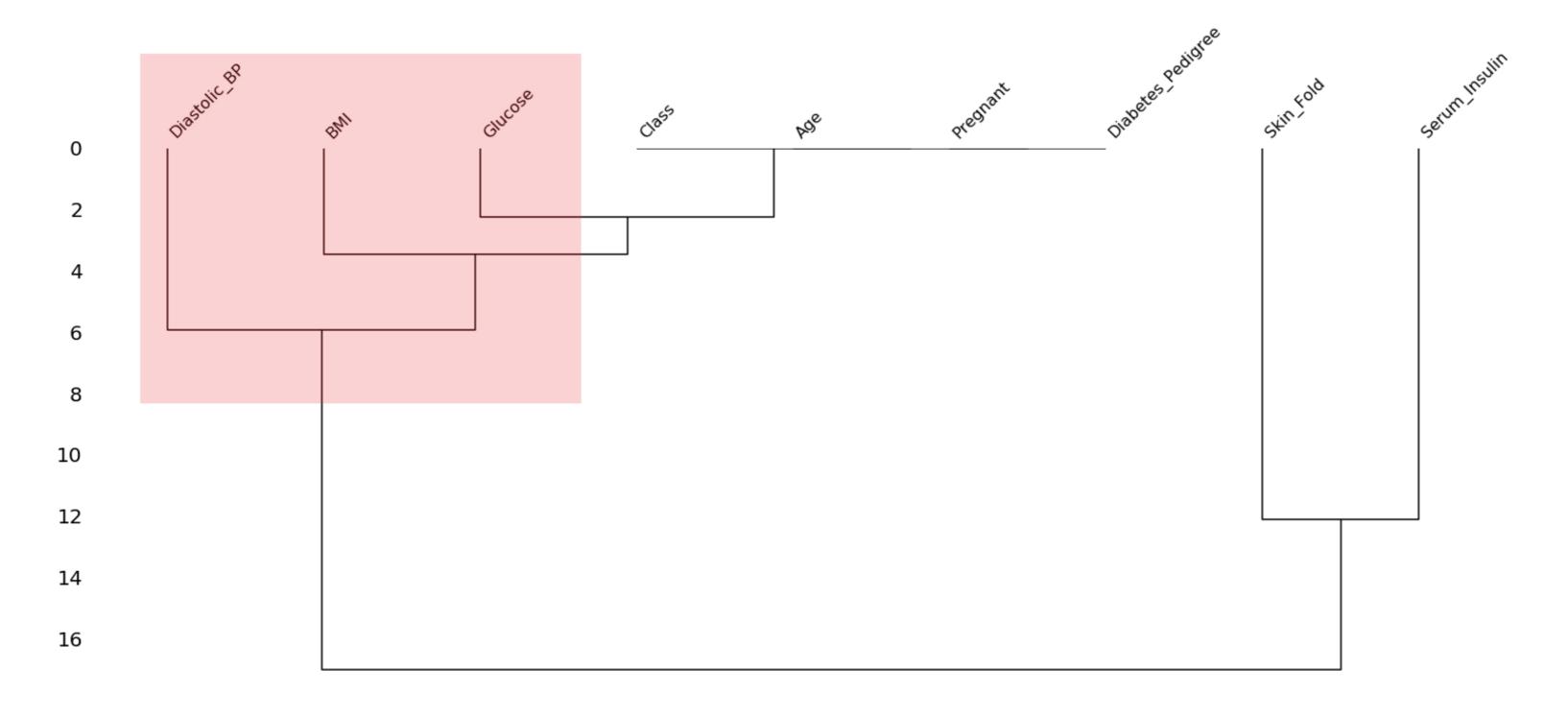
msno.dendrogram(diabetes)











Summary

Analyze missingness heatmap

```
msno.heatmap(df)
```

Analayze missingness dendrogram

```
msno.dendrogram(df)
```

Let's practice!

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Visualizing missingness across a variable

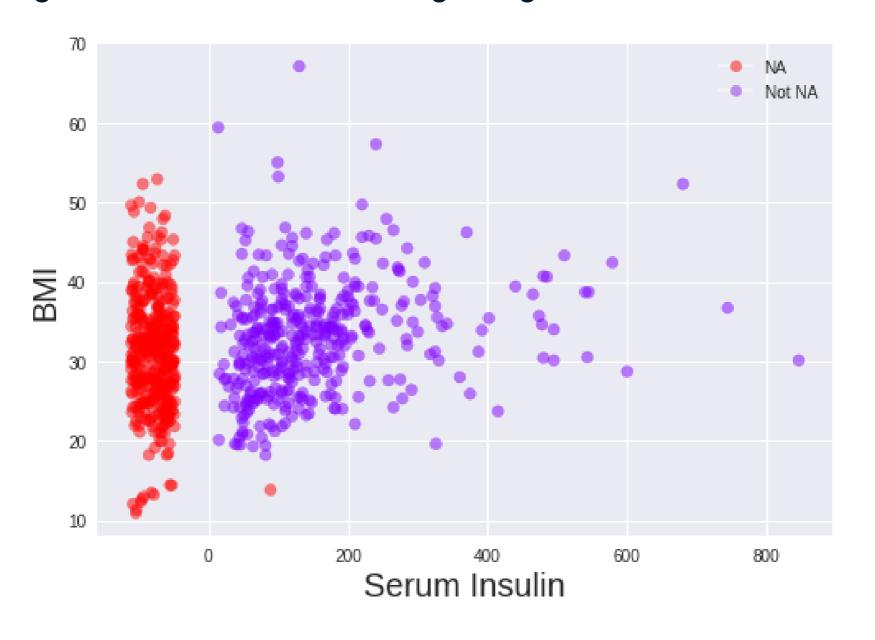
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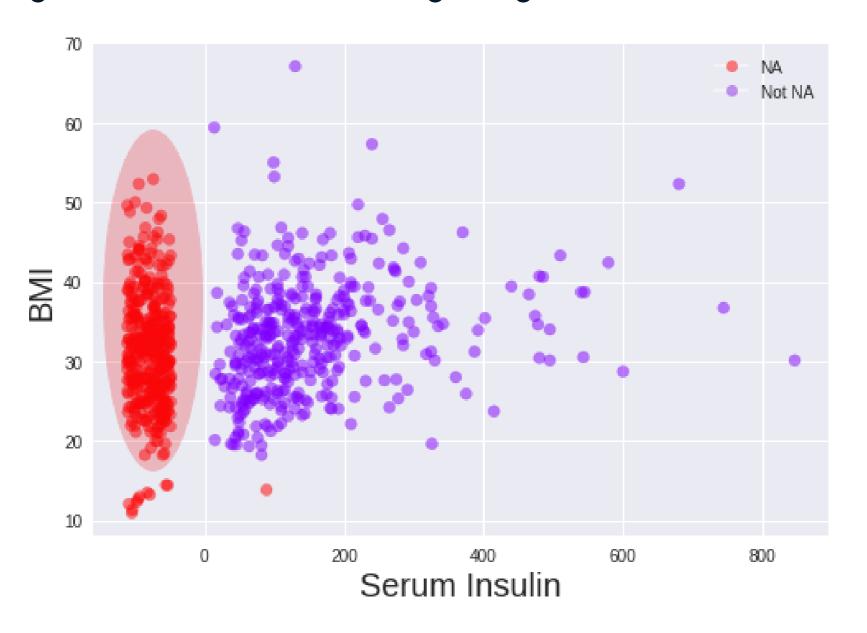
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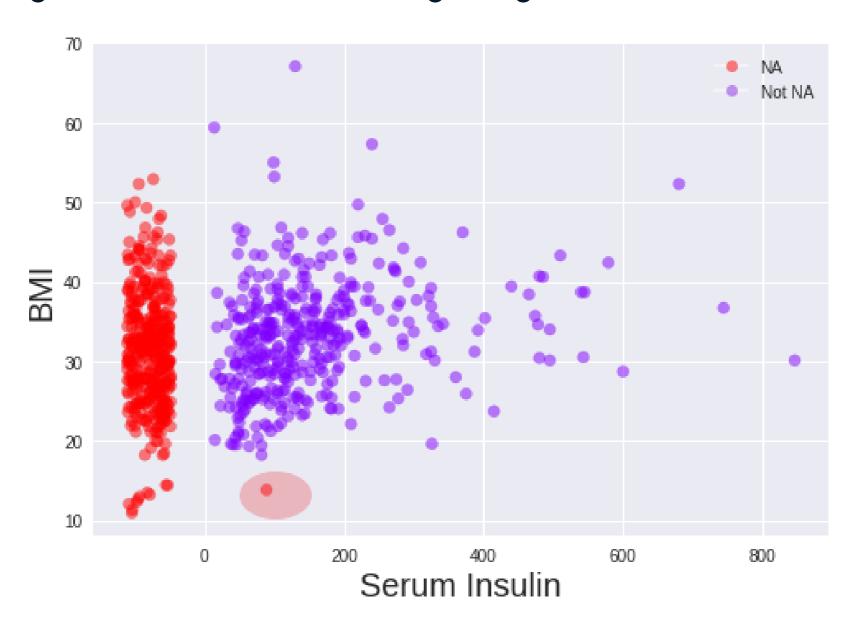
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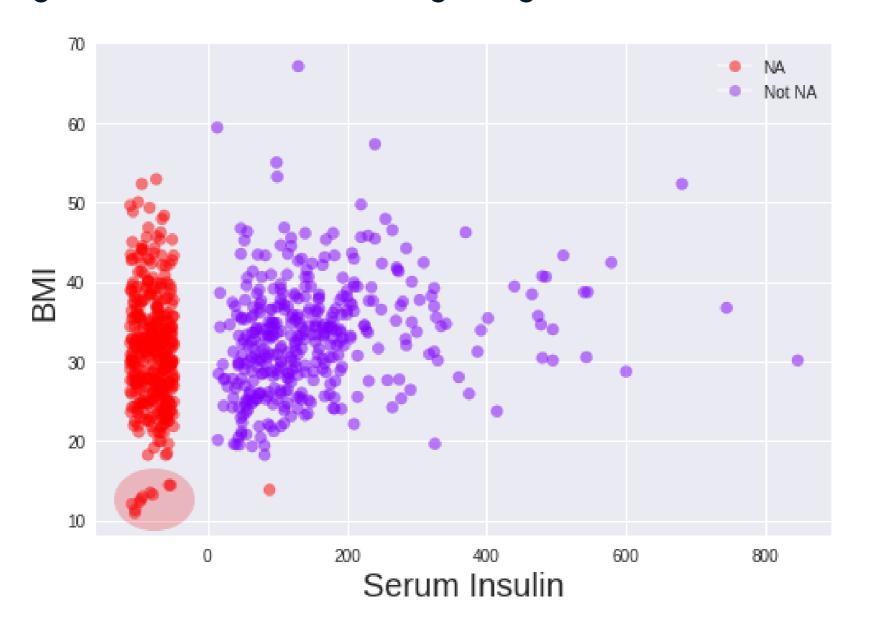


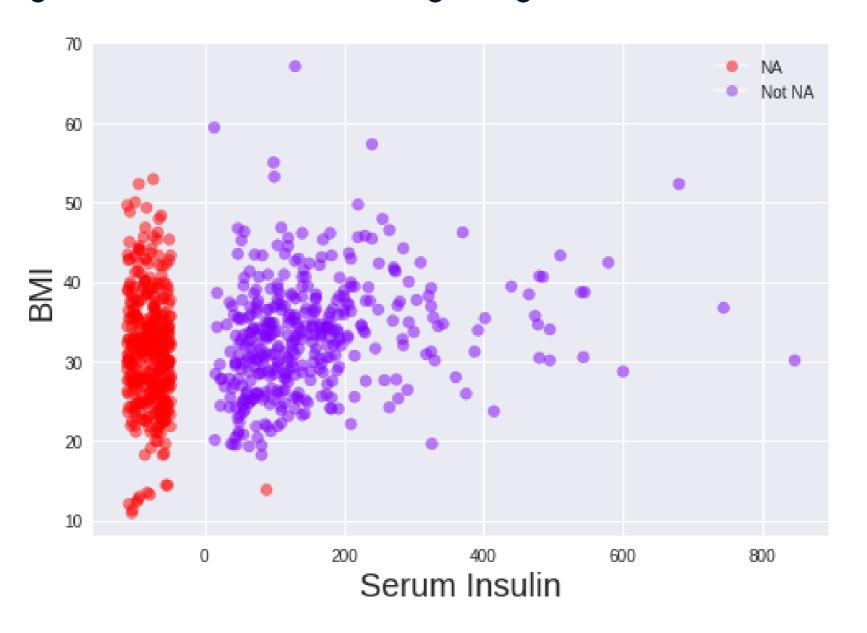








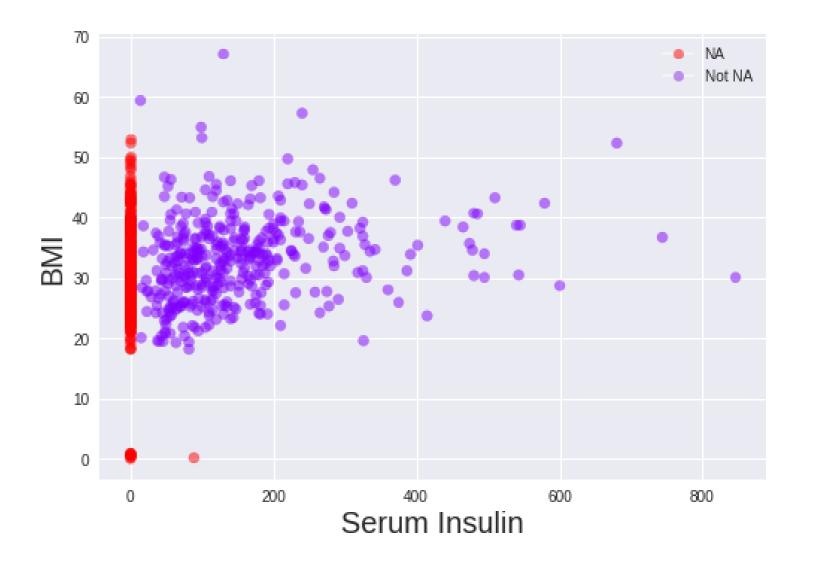




```
from numpy.random import rand

BMI_null = diabetes['BMI'].isnull()
num_nulls = BMI_null.sum()

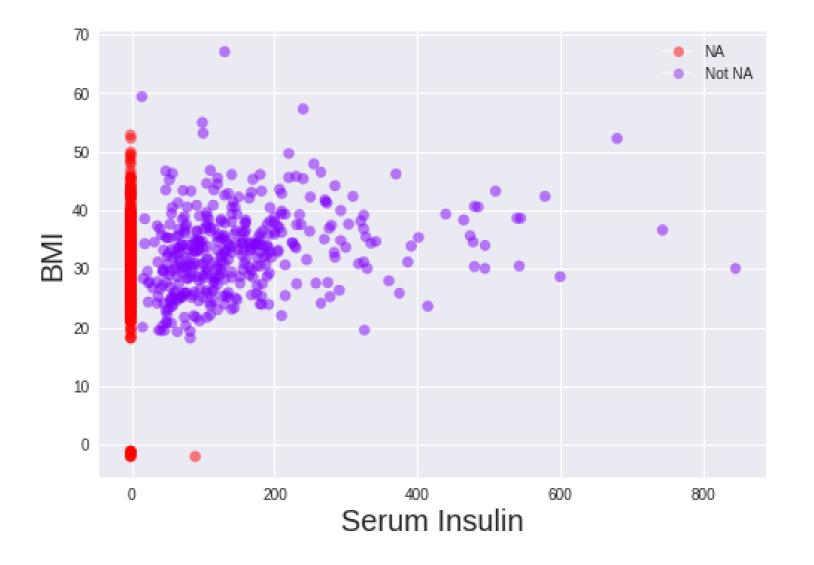
# Generate random values
dummy_values = rand(num_nulls)
```



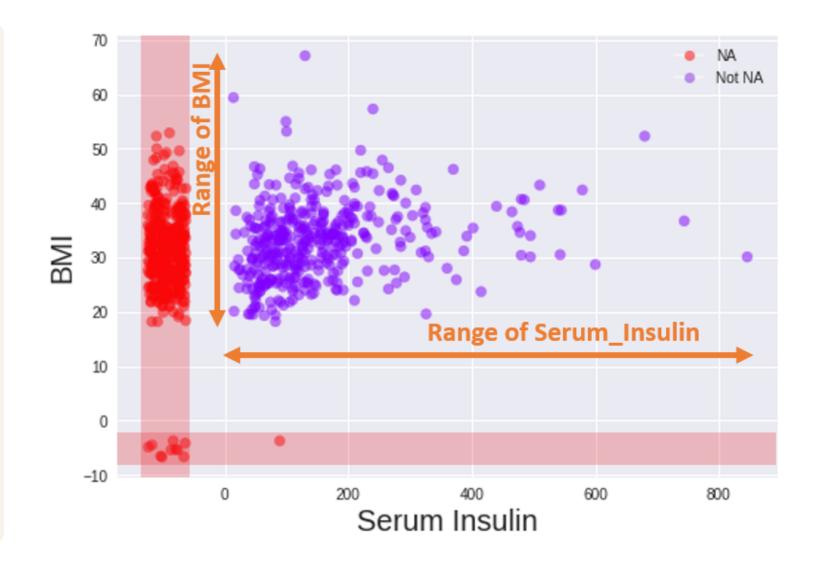
```
from numpy.random import rand

BMI_null = diabetes['BMI'].isnull()
num_nulls = BMI_null.sum()

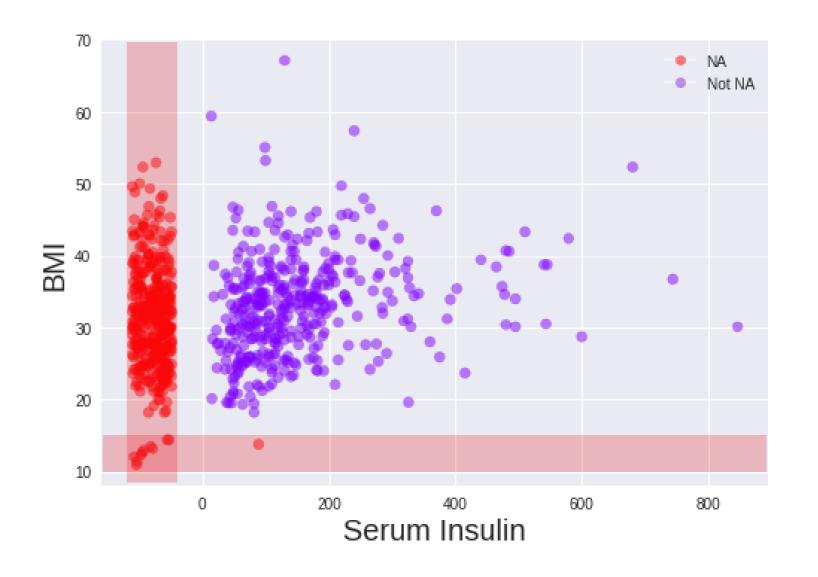
# Generate random values
dummy_values = rand(num_nulls)
# Shift to -2 & -1
dummy_values = dummy_values - 2
```



```
from numpy.random import rand
BMI_null = diabetes['BMI'].isnull()
num_nulls = BMI_null.sum()
# Generate random values
dummy_values = rand(num_nulls)
# Shift to -2 & -1
dummy_values = dummy_values - 2
# Scale to 0.075 of Column Range
BMI_range = BMI.max() - BMI.min()
dummy_values = dummy_values * 0.075 * BMI_range
```

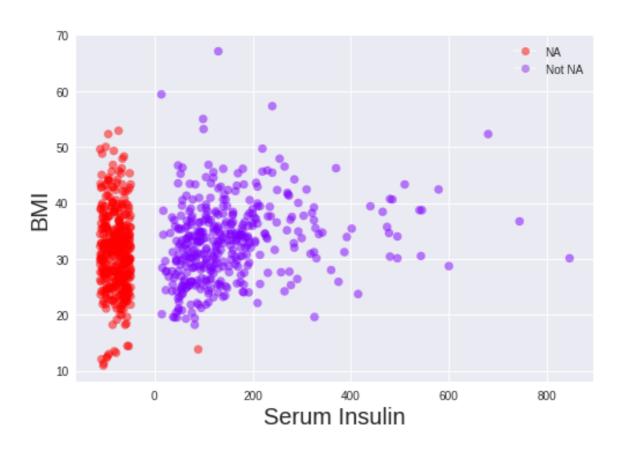


```
from numpy.random import rand
BMI_null = diabetes['BMI'].isnull()
num_nulls = BMI_null.sum()
# Generate random values
dummy_values = rand(num_nulls)
# Shift to -2 & -1
dummy_values = dummy_values - 2
# Scale to 0.075 of Column Range
BMI_range = BMI.max() - BMI.min()
dummy_values = dummy_values * 0.075 * BMI_range
# Shift to Column Minimum
dummy_values = (rand(num_nulls) - 2)
        * 0.075 * BMI_range + BMI.min()
```



```
from numpy.random import rand
def fill_dummy_values(df, scaling_factor):
  # Create copy of dataframe
  df_dummy = df.copy(deep=True)
  # Iterate over each column
  for col in df_dummy:
   # Get column, column missing values and range
    col = df_dummy[col]
    col_null = col.isnull()
    num_nulls = col_null.sum()
    col_range = col.max() - col.min()
   # Shift and scale dummy values
    dummy_values = (rand(num_nulls) - 2)
    dummy_values = dummy_values * scaling_factor * col_range + col.min()
   # Return dummy values
    col[col_null] = dummy_values
  return df_dummy
```





Let's practice!

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When and how to delete missing data

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Types of deletions

- 1. Pairwise deletion
- 2. Listwise deletion

Note: Used when the values are MCAR.

Pairwise Deletion

diabetes DataFrame

Pregnant	Glucose	Diastolic_BP	
6	148	72	•••
5	NaN	80	
1	89	66	•••
1	NaN	74	•••
			•••
8	183	64	
6	NaN	68	•••

768 rows × 9 columns

```
diabetes['Glucose'].mean()
121.687
diabetes.count()
763
diabetes['Glucose'].sum() /
       diabetes['Glucose'].count()
121.687
```

Listwise Deletion or Complete Case

diabetes DataFrame

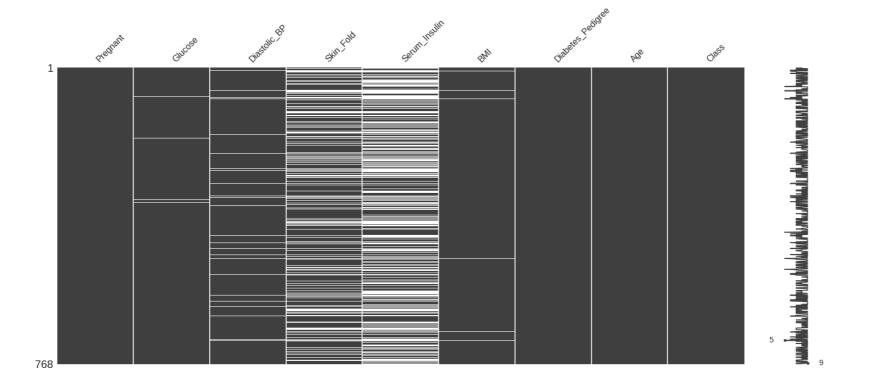
Pregnant	Glucose	Diastolic_BP	
6	148	72	
5	NaN	80	
1	89	66	
1	NaN	74	
	•••		•••
8	183	64	
6	NaN	68	

768 rows × 9 columns

Deletion in diabetes DataFrame

```
msno.matrix(diabetes)
diabetes['Glucose'].isnull().sum()
```

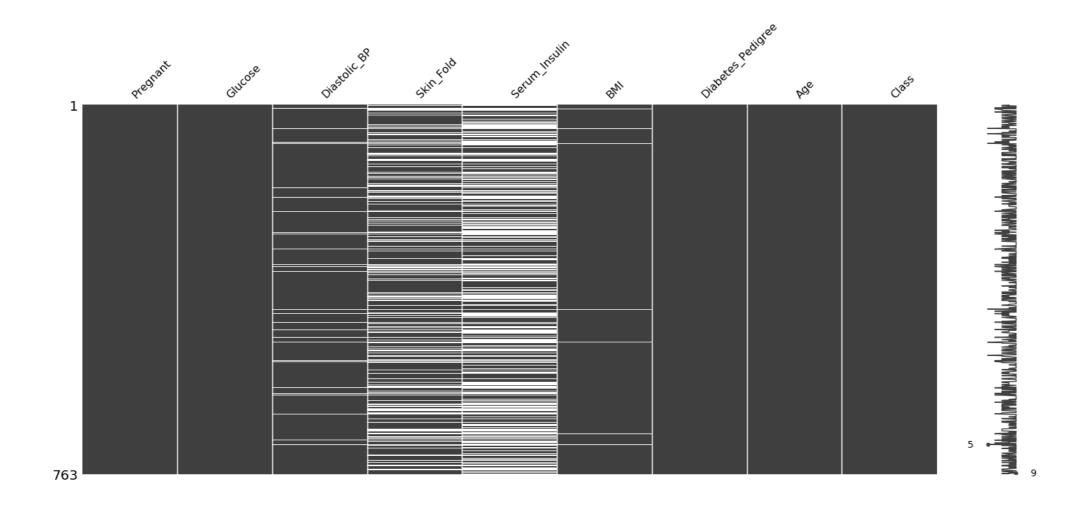
5





Deletion in diabetes DataFrame

```
diabetes.dropna(subset=["Glucose"], how='any', inplace=True)
msno.matrix(diabetes)
```



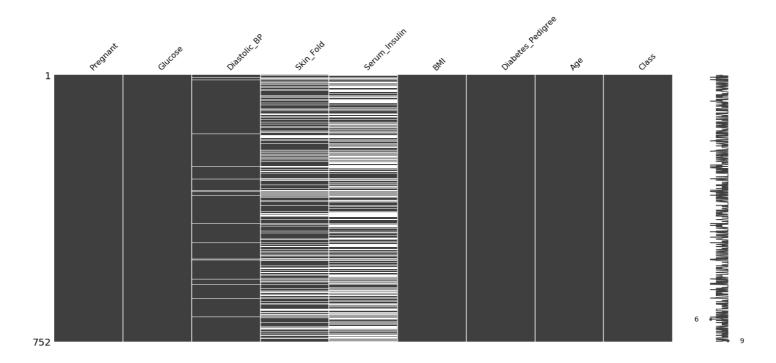


Deletion in diabetes DataFrame

```
diabetes['BMI'].isnull().sum()
```

11

```
diabetes.dropna(subset=["BMI"], how='any', inplace=True)
msno.matrix(diabetes)
```





Summary

- Pairwise deletion
- Listwise deletion
- Deletion is used only when values are MCAR

Let's practice!

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