

In [6]:

```
import sys
import numpy
import matplotlib
import pandas
import sklearn

print('Python: {}'.format(sys.version))
print('numpy: {}'.format(numpy.__version__))
print('matplotlib: {}'.format(matplotlib.__version__))
print('pandas: {}'.format(pandas.__version__))
print('sklearn: {}'.format(sklearn.__version__))
```

Python: 3.7.6 (default, Jan 8 2020, 20:23:39) [MSC v.1916 64 bit (AMD64)]  
numpy: 1.18.1  
matplotlib: 3.1.3  
pandas: 1.0.1  
sklearn: 0.22.1

In [8]:

```
import numpy as np
from sklearn import preprocessing
from sklearn.model_selection import cross_validate
from sklearn.neighbors import KNeighborsClassifier
from sklearn.svm import SVC
from sklearn import model_selection
from sklearn.metrics import classification_report, accuracy_score
from pandas.plotting import scatter_matrix
import matplotlib.pyplot as plt
import pandas as pd
```

In [24]:

```
#url="E:\ML\Projects\Breast Cancer Detection\breast-cancer-wisconsin.data"
url = "https://archive.ics.uci.edu/ml/machine-learning-databases/breast-cancer-wisconsin/breast-cancer-wisconsin.data"
names=['id', 'clump_thickness', 'uniform_cell_size', 'uniform_cell_shape',
       'marginal_adhesion', 'single_epithelial_size', 'bare_nuclei',
       'bland_chromatin', 'normal_nucleoli', 'mitoses', 'class']
df = pd.read_csv(url, names=names)
```

In [25]:

```
df.replace('?', -99999, inplace=True)
print(df.axes)
```

```
[RangeIndex(start=0, stop=699, step=1), Index(['id', 'clump_thickness', 'uniform_cell_size', 'uniform_cell_shape',
       'marginal_adhesion', 'single_epithelial_size', 'bare_nuclei',
       'bland_chromatin', 'normal_nucleoli', 'mitoses', 'class'],
       dtype='object')]
```

In [26]:

```
#print(df)
#print(df.shape)
df.drop(['id'],1,inplace=True)
print(df)
print(df.shape)
```

|     | clump_thickness | uniform_cell_size | uniform_cell_shape | \ |
|-----|-----------------|-------------------|--------------------|---|
| 0   | 5               | 1                 | 1                  |   |
| 1   | 5               | 4                 | 4                  |   |
| 2   | 3               | 1                 | 1                  |   |
| 3   | 6               | 8                 | 8                  |   |
| 4   | 4               | 1                 | 1                  |   |
| ..  | ...             | ...               | ...                |   |
| 694 | 3               | 1                 | 1                  |   |
| 695 | 2               | 1                 | 1                  |   |
| 696 | 5               | 10                | 10                 |   |
| 697 | 4               | 8                 | 6                  |   |
| 698 | 4               | 8                 | 8                  |   |

|     | marginal_adhesion | single_epithelial_size | bare_nuclei | bland_chromati |
|-----|-------------------|------------------------|-------------|----------------|
| n \ |                   |                        |             |                |
| 0   | 1                 | 2                      | 1           |                |
| 3   |                   |                        |             |                |
| 1   | 5                 | 7                      | 10          |                |
| 3   |                   |                        |             |                |
| 2   | 1                 | 2                      | 2           |                |
| 3   |                   |                        |             |                |
| 3   | 1                 | 3                      | 4           |                |
| 3   |                   |                        |             |                |
| 4   | 3                 | 2                      | 1           |                |
| 3   |                   |                        |             |                |
| ..  | ...               | ...                    | ...         |                |
| ... |                   |                        |             |                |
| 694 | 1                 | 3                      | 2           |                |
| 1   |                   |                        |             |                |
| 695 | 1                 | 2                      | 1           |                |
| 1   |                   |                        |             |                |
| 696 | 3                 | 7                      | 3           |                |
| 8   |                   |                        |             |                |
| 697 | 4                 | 3                      | 4           | 1              |
| 0   |                   |                        |             |                |
| 698 | 5                 | 4                      | 5           | 1              |
| 0   |                   |                        |             |                |

|     | normal_nucleoli | mitoses | class |
|-----|-----------------|---------|-------|
| 0   | 1               | 1       | 2     |
| 1   | 2               | 1       | 2     |
| 2   | 1               | 1       | 2     |
| 3   | 7               | 1       | 2     |
| 4   | 1               | 1       | 2     |
| ..  | ...             | ...     | ...   |
| 694 | 1               | 1       | 2     |
| 695 | 1               | 1       | 2     |
| 696 | 10              | 2       | 4     |
| 697 | 6               | 1       | 4     |
| 698 | 4               | 1       | 4     |

[699 rows x 10 columns]  
(699, 10)

In [28]:

```
print(df.loc[6])
```

```
clump_thickness      1
uniform_cell_size    1
uniform_cell_shape    1
marginal_adhesion    1
single_epithelial_size 2
bare_nuclei          10
bland_chromatin       3
normal_nucleoli       1
mitoses              1
class                2
Name: 6, dtype: object
```

In [29]:

```
print(df.describe())
```

|       | clump_thickness | uniform_cell_size | uniform_cell_shape \ |
|-------|-----------------|-------------------|----------------------|
| count | 699.000000      | 699.000000        | 699.000000           |
| mean  | 4.417740        | 3.134478          | 3.207439             |
| std   | 2.815741        | 3.051459          | 2.971913             |
| min   | 1.000000        | 1.000000          | 1.000000             |
| 25%   | 2.000000        | 1.000000          | 1.000000             |
| 50%   | 4.000000        | 1.000000          | 1.000000             |
| 75%   | 6.000000        | 5.000000          | 5.000000             |
| max   | 10.000000       | 10.000000         | 10.000000            |

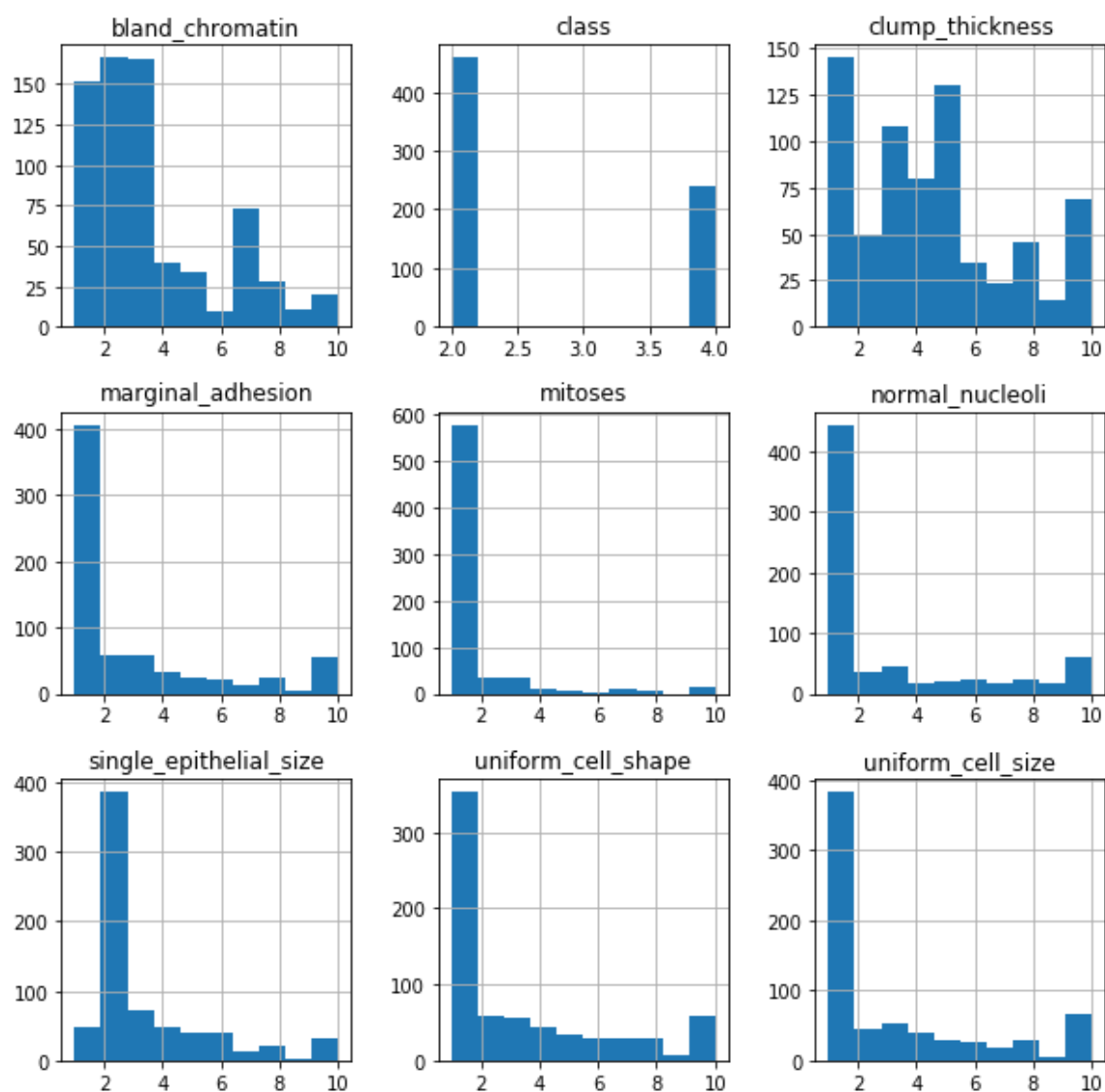
|       | marginal_adhesion | single_epithelial_size | bland_chromatin \ |
|-------|-------------------|------------------------|-------------------|
| count | 699.000000        | 699.000000             | 699.000000        |
| mean  | 2.806867          | 3.216023               | 3.437768          |
| std   | 2.855379          | 2.214300               | 2.438364          |
| min   | 1.000000          | 1.000000               | 1.000000          |
| 25%   | 1.000000          | 2.000000               | 2.000000          |
| 50%   | 1.000000          | 2.000000               | 3.000000          |
| 75%   | 4.000000          | 4.000000               | 5.000000          |
| max   | 10.000000         | 10.000000              | 10.000000         |

|       | normal_nucleoli | mitoses    | class      |
|-------|-----------------|------------|------------|
| count | 699.000000      | 699.000000 | 699.000000 |
| mean  | 2.866953        | 1.589413   | 2.689557   |
| std   | 3.053634        | 1.715078   | 0.951273   |
| min   | 1.000000        | 1.000000   | 2.000000   |
| 25%   | 1.000000        | 1.000000   | 2.000000   |
| 50%   | 1.000000        | 1.000000   | 2.000000   |
| 75%   | 4.000000        | 1.000000   | 4.000000   |
| max   | 10.000000       | 10.000000  | 4.000000   |

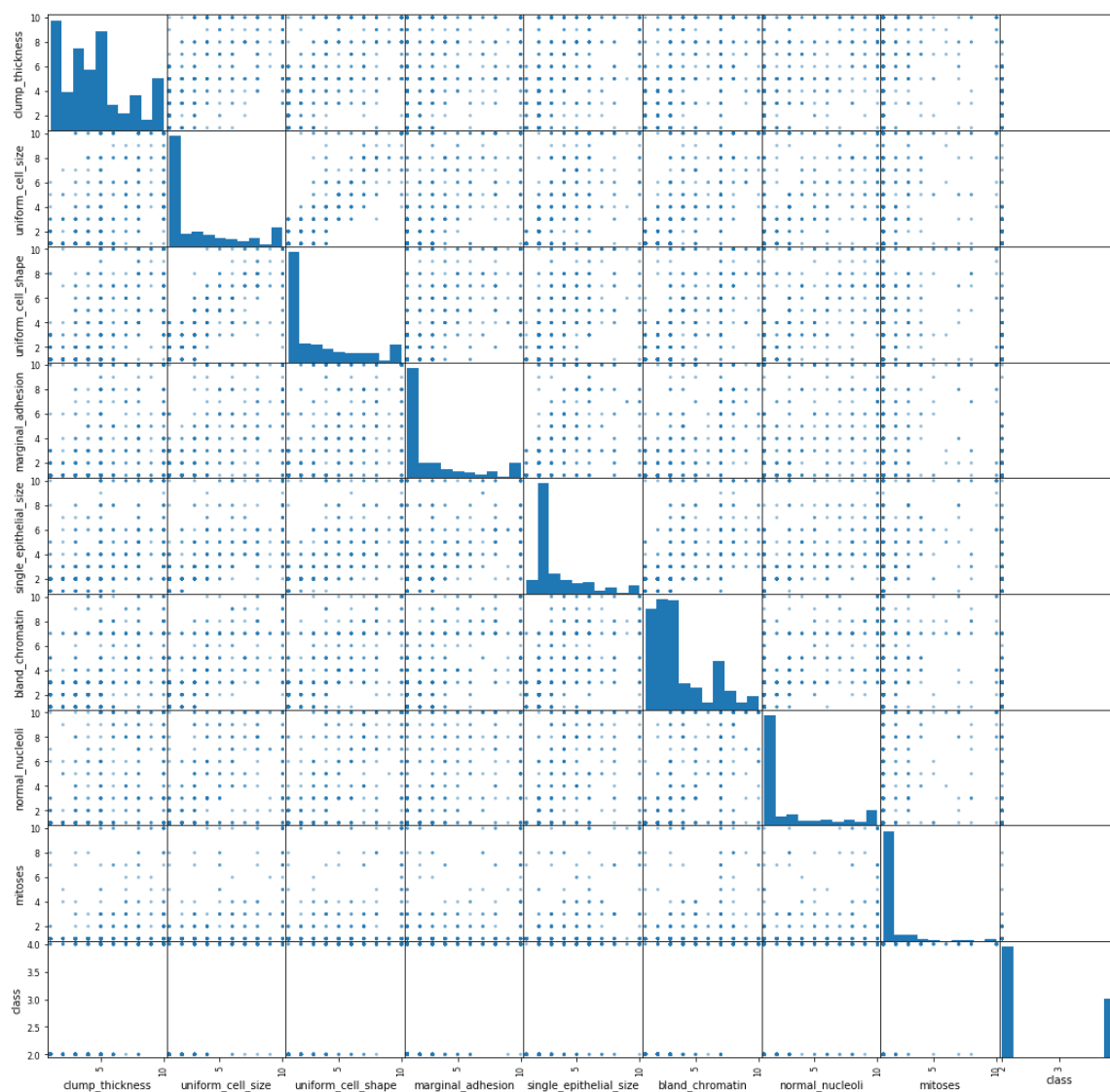
In [31]:

```
df.hist(figsize = (10, 10))  
plt.show()
```



In [210]:

```
scatter_matrix(df,figsize=(18,18))  
plt.show()
```



In [500]:

```

X=np.array(df.drop(['class'],1))
y=np.array(df['class'])
X_train, X_test, y_train, y_test = model_selection.train_test_split(X, y, test_size=0.2
)

models=[]
models.append(('KNN',KNeighborsClassifier(n_neighbors=6)))
models.append(('SVM',SVC(kernel='linear')))
#models.append(('SVM',SVC(kernel='rbf')))
#models.append(('SVM',SVC(kernel='poly')))
#models.append(('SVM',SVC(kernel='sigmoid')))

results = []
names = []

for name,model in models:
    kfold=model_selection.KFold(n_splits=10, random_state = 0, shuffle=True)
    cv_results=model_selection.cross_val_score(model,X_train,y_train,cv=kfold,scoring=
'accuracy')
    results.append(cv_results)
    names.append(name)
    print("{}:accuracy->{}(std->{})".format(name, cv_results.mean(), cv_results.std()))

```

KNN:accuracy-&gt;0.9678571428571427(std-&gt;0.022303564279994276)

SVM:accuracy-&gt;0.9446103896103896(std-&gt;0.03780654211128742)

In [501]:

```

for name, model in models:
    model.fit(X_train, y_train)
    predictions = model.predict(X_test)
    print(name)
    print(accuracy_score(y_test, predictions))
    print(classification_report(y_test, predictions))

```

KNN

0.9714285714285714

|              | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 2            | 0.99      | 0.97   | 0.98     | 97      |
| 4            | 0.93      | 0.98   | 0.95     | 43      |
| accuracy     |           |        | 0.97     | 140     |
| macro avg    | 0.96      | 0.97   | 0.97     | 140     |
| weighted avg | 0.97      | 0.97   | 0.97     | 140     |

SVM

0.95

|              | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 2            | 0.99      | 0.94   | 0.96     | 97      |
| 4            | 0.88      | 0.98   | 0.92     | 43      |
| accuracy     |           |        | 0.95     | 140     |
| macro avg    | 0.93      | 0.96   | 0.94     | 140     |
| weighted avg | 0.95      | 0.95   | 0.95     | 140     |

In [508]:

```
clf = SVC(kernel='linear')

clf.fit(X_train, y_train)
accuracy = clf.score(X_test, y_test)
print(accuracy)

example_measures = np.array([[4,1,3,5,2,1,4,1,10]])
#print(example_measures)
example_measures = example_measures.reshape(len(example_measures), -1)
#print(example_measures)
prediction = clf.predict(example_measures)
print(prediction)
```

0.95

[4]