# In [52]:

```
import pandas
from sklearn.model_selection import train_test_split
from sklearn.naive_bayes import GaussianNB, BernoulliNB, MultinomialNB
import numpy as np
from sklearn import metrics
import matplotlib.pyplot as plt
```

## In [4]:

```
df = pandas.read_csv("train.csv")
```

## In [7]:

df.head()

## Out[7]:

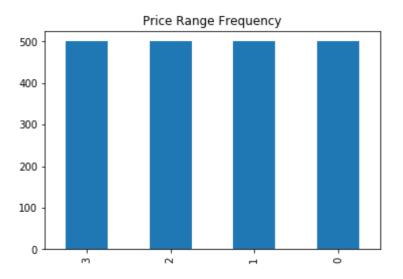
speed	dual_sim	fc	four_g	int_memory	m_dep	mobile_wt	n_cores	 px_height	px_width
2.2	0	1	0	7	0.6	188	2	 20	756
0.5	1	0	1	53	0.7	136	3	 905	1988
0.5	1	2	1	41	0.9	145	5	 1263	1716
2.5	0	0	0	10	8.0	131	6	 1216	1786
1.2	0	13	1	44	0.6	141	2	 1208	1212

## In [20]:

df.price\_range.value\_counts().plot(kind='bar', title='Price Range Frequency')

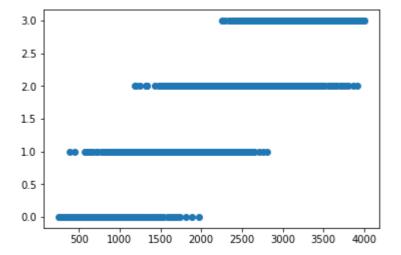
# Out[20]:

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f61ad6c5ba8>



```
In [62]:
```

```
plt.scatter(x=df['ram'], y=df['price_range'])
plt.show()
```



### In [ ]:

```
plt.scatter(x=df['ram'], y=df['price_range'])
plt.show()
```

#### In [42]:

```
#Dataset is pretty balanced, dividing the features_target dataset
target = np.array(df[["price_range"]].copy())
features = np.array(df.drop('price_range', axis = 1))
```

### In [43]:

```
#Dividing the train/test dataset
train_features, test_features, train_labels, test_labels = train_test_split(features, targetest)
```

## In [46]:

```
model = GaussianNB()
model.fit(train_features, train_labels.ravel())
```

### Out[46]:

GaussianNB(priors=None, var\_smoothing=1e-09)

#### In [47]:

```
y_pred = model.predict(test_features)
```

```
In [51]:
```

```
print('Expected performance')
print('Mean Absolute Error:', metrics.mean_absolute_error(test_labels, y_pred))
print('Mean Squared Error:', metrics.mean_squared_error(test_labels, y_pred))
print('Root Mean Squared Error:', np.sqrt(metrics.mean_squared_error(test_labels, y_pred)))
Expected performance
Mean Absolute Error: 0.208
Mean Squared Error: 0.208
Root Mean Squared Error: 0.45607017003965516
In [57]:
errors = 0
for i in range(len(test_labels)):
    if test_labels[i] != y_pred[i]:
        errors += 1
In [59]:
print("Naive Bayes hit rate: "+ str((len(test_labels)-errors)/(len(test_labels)) * 100) +
Naive Bayes hit rate: 79.2%
In [ ]:
```