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**Materia: Simulación**

## REGESION LINEAL

In [36]:

```
1 import numpy as np
2 import random
3 from sklearn import linear_model
4 from sklearn.metrics import mean_squared_error, r2_score
5 import matplotlib.pyplot as plt
6 %matplotlib inline
7
```

In [37]:

```
1 # Generador de distribución de datos para regresión lineal simple
2 def generador_datos_simple(beta, muestras, desviacion):
3     # Genero n (muestras) valores de x aleatorios entre 0 y 100
4     x = np.random.random(muestras) * 100
5     # Genero un error aleatorio gaussiano con desviación típica (desviacion)
6     e = np.random.randn(muestras) * desviacion
7     # Obtengo el y real como x*beta + error
8     y = x * beta + e
9     return x.reshape((muestras,1)), y.reshape((muestras,1))
10
```

In [43]:

```
1 # Parámetros de la distribución
2 desviacion = 200
3 beta = 10
4 n = 50
5 x, y = generador_datos_simple(beta, n, desviacion)
6 print('Datos en X:',x)
7 print('Datos en Y:',y)
```

Datos en X: [[44.91288826]

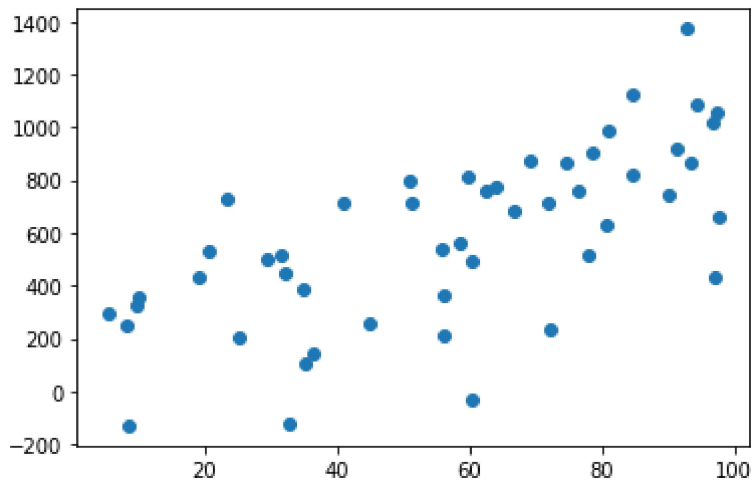
[82.30395836]  
[58.29539527]  
[40.58721592]  
[30.93105732]  
[98.81748645]  
[19.35627593]  
[72.35250848]  
[60.38276856]  
[13.03075465]  
[76.77210683]  
[24.89638097]  
[61.18005996]  
[ 8.45011698]  
[38.56447357]  
[77.44326889]  
[27.8872075 ]  
[70.8695727 ]  
[65.34632199]  
[12.90286206]  
[72.6620095 ]  
[76.90926054]  
[19.81479127]  
[62.05090138]  
[95.9187143 ]  
[ 4.56228532]  
[ 1.88160931]  
[ 6.62787421]  
[41.49511413]  
[ 2.52654166]  
[32.70465605]  
[ 8.56434888]  
[74.13361337]  
[74.05935032]  
[10.13079302]  
[51.41083397]  
[50.82729727]  
[80.65312877]  
[53.17497356]  
[81.31229518]  
[ 9.07797807]  
[18.64701846]  
[25.13711991]  
[87.84611229]  
[96.57979064]  
[19.48309584]  
[ 2.96616018]  
[27.88893841]  
[21.08646798]  
[33.70436481]]

Datos en Y: [[ 509.23170814]

```
[ 793.15985488]
[ 768.93564387]
[ 384.10464493]
[ 125.60904219]
[ 892.65841346]
[  80.24579104]
[ 711.25865055]
[ 659.82663121]
[ -99.94912274]
[ 919.79572091]
[ 594.21605152]
[ 872.78485601]
[-235.6254177 ]
[ 468.77224523]
[ 875.40249712]
[  60.40472948]
[ 357.5686108 ]
[ 525.96923879]
[ 263.97403018]
[ 400.15031053]
[ 353.91286291]
[ 277.13021937]
[ 785.85749829]
[1057.92106267]
[ -82.11387211]
[-120.02491297]
[ -26.13086175]
[ 614.81808914]
[ 152.83407715]
[ 499.78092293]
[ 102.86204477]
[ 879.22189549]
[ 976.67062   ]
[ -5.3985307  ]
[ 574.54884925]
[ 281.58990561]
[ 543.59187891]
[ 628.44098384]
[ 791.70644438]
[ 114.10533045]
[  74.01343406]
[ 372.97047279]
[1161.44204856]
[ 616.77532454]
[ 242.70923053]
[ 307.27729361]
[ 756.10718363]
[ 405.79927587]
[ 443.93957805]]
```

In [39]:

```
1 # Represento Los datos generados
2 plt.scatter(x, y)
3 plt.show()
4
```



In [40]:

```
1 # Creo un modelo de regresión Lineal
2 modelo = linear_model.LinearRegression()
3
4 # Entreno el modelo con los datos (X,Y)
5 modelo.fit(x, y)
6 # Ahora puedo obtener el coeficiente b_1
7 print(u'Coeficiente de determinación: ', modelo.coef_[0])
8
9 # Podemos predecir usando el modelo
10 y_pred = modelo.predict(x)
11
12 # Por último, calculamos el error cuadrático medio y el estadístico R^2
13 print(u'Error cuadrático medio: %.2f' % mean_squared_error(y, y_pred))
14 print(u'Estadístico R_2: %.2f' % r2_score(y, y_pred))
15
```

Coeficiente beta1: [7.89148096]  
Error cuadrático medio: 59266.78  
Estadístico R\_2: 0.45

In [41]:

```
1 # Representamos el ajuste (rojo) y la recta  $Y = \text{beta} * x$  (verde)
2 plt.scatter(x, y)
3 plt.plot(x, y_pred, color='red')
4 x_real = np.array([0, 100])
5 y_real = x_real * beta
6 plt.plot(x_real, y_real, color='green')
7 plt.show()
8
```

