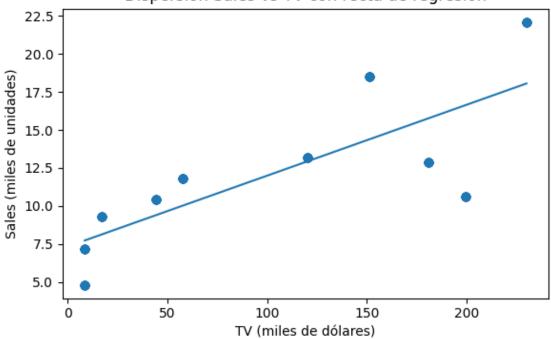
Tarea2

August 21, 2025

```
[]: import pandas as pd
    import matplotlib.pyplot as plt
    import numpy as np
    df = pd.read_csv("Advertising.csv")
    df.head()
[]:
          TV Radio Newspaper Sales
    0 230.1
               37.8
                          69.2
                                 22.1
    1 44.5
               39.3
                          45.1
                                 10.4
    2 17.2
               45.9
                          69.3
                                  9.3
    3 151.5
               41.3
                          58.5
                                 18.5
    4 180.8
               10.8
                          58.4
                                 12.9
[2]: # Variable independiente X (TV) y dependiente Y (Sales)
    x = df["TV"].astype(float).values
    y = df["Sales"].astype(float).values
    n = len(x)
    x_{mean} = x.mean()
    y_mean = y.mean()
    n, x_mean, y_mean
[2]: (200, np.float64(101.89), np.float64(12.08))
[3]: # Cálculo de la recta por fórmulas (sin statsmodels)
    Sxx = np.sum((x - x_mean)**2)
    Sxy = np.sum((x - x_mean)*(y - y_mean))
    beta1 = Sxy / Sxx
    beta0 = y_mean - beta1 * x_mean
    # Gráfico
    plt.figure(figsize=(6,4))
    plt.scatter(x, y)
    x_line = np.linspace(x.min(), x.max(), 100)
    y_line = beta0 + beta1 * x_line
    plt.plot(x_line, y_line)
    plt.xlabel("TV (miles de dólares)")
```

```
plt.ylabel("Sales (miles de unidades)")
plt.title("Dispersión Sales vs TV con recta de regresión")
plt.tight_layout()
plt.show()
```

Dispersión Sales vs TV con recta de regresión



```
[4]: # Cálculo manual equivalente
Syy = np.sum((y - y_mean)**2)
r = Sxy / np.sqrt(Sxx * Syy)
r
```

[4]: np.float64(0.7769242755776288)

```
[5]: beta0, beta1
```

[5]: (np.float64(7.334018424710768), np.float64(0.04657946388545718))

```
[7]: y_hat = beta0 + beta1 * x
resid = y - y_hat

SSE = np.sum(resid**2)  # Sum of Squared Errors
SST = np.sum((y - y_mean)**2)  # Total Sum of Squares
SSR = SST - SSE  # Regression Sum of Squares
```

```
R2 = SSR / SST

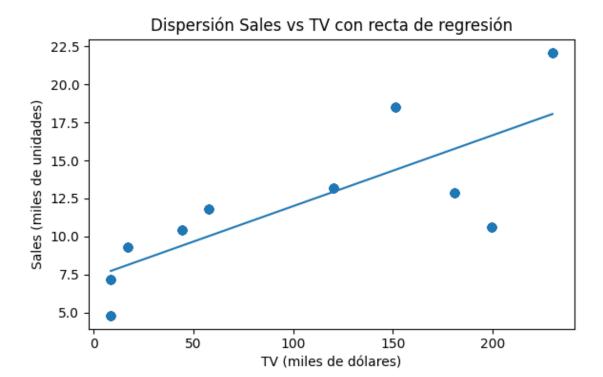
MSE = SSE / (n - 2) # Error cuadrático medio (estimado con df = \frac{1}{2} + \frac{1}{2} R2, MSE
```

[7]: (np.float64(0.6036113299818231), np.float64(9.3361943959756))

```
[8]: df_reg = 1
     df_res = n - 2
     df_{tot} = n - 1
     MSR = SSR / df_reg
     MSE = SSE / df_res
     F = MSR / MSE
     anova = pd.DataFrame({
         "Fuente": ["Regresión", "Residual", "Total"],
                    [df_reg, df_res, df_tot],
         "SS":
                    [SSR,
                              SSE,
                                      SST],
                                      ""],
         "MS":
                    [MSR,
                              MSE,
                              0.0
                                       "" ]
         ^{0}F^{0}:
                    [F,
     })
     anova
```

```
[8]: Fuente df SS MS F
0 Regresión 1 2814.95351 2814.95351 301.509736
1 Residual 198 1848.56649 9.336194
2 Total 199 4663.52000
```

```
[9]: plt.figure(figsize=(6,4))
   plt.scatter(x, y)
   plt.plot(x_line, y_line)
   plt.xlabel("TV (miles de dólares)")
   plt.ylabel("Sales (miles de unidades)")
   plt.title("Dispersión Sales vs TV con recta de regresión")
   plt.tight_layout()
   plt.savefig("scatter_tv_sales.png", dpi=200)
```



```
[10]: print(f"n = {n}")
    print(f"Modelo: Sales = {beta0:.4f} + {beta1:.4f} \cdot TV")
    print(f"r = {r:.4f}")
    print(f"R^2 = {R2:.4f}")
    print(f"SST = {SST:.6f} | SSR = {SSR:.6f} | SSE = {SSE:.6f}")
    print(f"MSE = {MSE:.6f}")
    print(f"F = {F:.4f}")

n = 200
    Modelo: Sales = 7.3340 + 0.0466 \cdot TV
    r = 0.7769
    R^2 = 0.6036
    SST = 4663.520000 | SSR = 2814.953510 | SSE = 1848.566490
    MSE = 9.336194
    F = 301.5097
```