10_regression_confidence_regions

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[]: import os
user = os.getenv('USER')
os.chdir(f'/scratch/cd82/{user}/notebooks')
```

1 Linear Regression - Confidence region of regression line

Confidence intervals on estimates $\,$ For a choosen significance level, with degrees of freedom =n-2

$$CSS_{x} = \sum_{i=1}^{n} x^{2} - \frac{(\sum_{i=1}^{n} x)^{2}}{n}$$

$$\Delta\beta_1 = \pm t_{(\alpha/2, n-2)} \sqrt{\frac{s^2}{CSS_x}}$$

$$\Delta\beta_0 = \pm t_{(\alpha/2,n-2)} \sqrt{\frac{s^2 \cdot \sum_{i=1}^n x^2}{n \cdot CSS_x}}$$

- CSS_x is the corrected sum of squares
- $\Delta \beta_1$ is the error range of the slope
- $\Delta\beta_0$ is the error range of the intercept
- s^2 is the variance

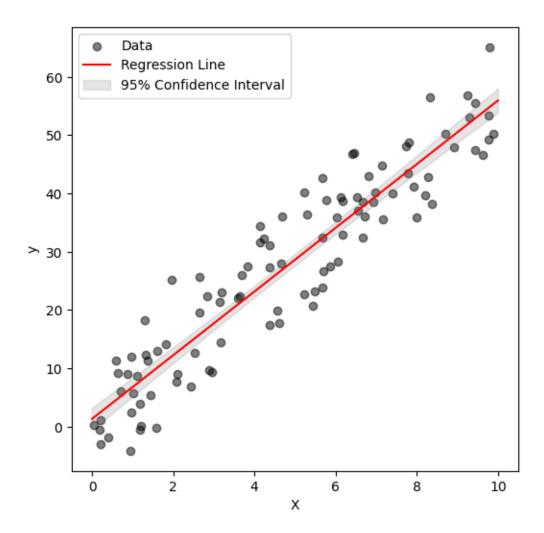
```
[1]: import numpy as np
  import matplotlib.pyplot as plt
  from sklearn.linear_model import LinearRegression
  from sklearn.model_selection import train_test_split
  from sklearn.metrics import mean_squared_error, r2_score
  from scipy import stats
```

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[8]: # Sample data
N=100
np.random.seed(0)
X = np.random.rand(N, 1) * 10
y = 5.5 * X.squeeze() + np.random.randn(N) * 6.0

# Fit the linear regression model
model = LinearRegression()
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model.fit(X, y)
# Predictions
X_{pred} = np.linspace(0, 10, 100).reshape(-1, 1)
y_pred = model.predict(X_pred)
# The two tail significance
significance = 0.95
sign_2tail = (1.0 - significance) / 2.0
sign_2tail = significance + sign_2tail
print('significance =', sign_2tail)
# Calculate the confidence intervals
n = N \# len(X)
mse = np.mean((y - model.predict(X))**2)
print('mse.shape: ', mse.shape)
t_val = stats.t.ppf(significance, n - 2) # 95% confidence interval in 2-tail_
se = np.sqrt(mse * (1/n + (X_pred - np.mean(X))**2 / np.sum((X - np.
 \rightarrowmean(X))**2)))
ci = t_val * se
# Plotting
plt.figure(figsize=(6, 6))
plt.scatter(X, y, color='black', alpha=0.5, label='Data')
plt.plot(X_pred, y_pred, color='red', label='Regression Line')
plt.fill_between(X_pred.squeeze(), (y_pred - ci.T).T.squeeze(), (y_pred + ci.T).
 ⇔T.squeeze(),
                 color='gray', alpha=0.2, label='95% Confidence Interval')
plt.xlabel('X')
plt.ylabel('y')
plt.legend()
plt.show()
```

significance = 0.975
mse.shape: ()



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