

In a lab experiment the following values are measured:

$\Psi / ^\circ$	Asymmetry	$\Psi / ^\circ$	Asymmetry	$\Psi / ^\circ$	Asymmetry
0	-0.032	30	0.010	60	0.057
90	0.068	120	0.076	150	0.080
180	0.031	210	0.005	240	-0.041
270	-0.090	300	-0.088	330	-0.074

The asymmetry values have a measurement value of ± 0.011 . The theory says that the asymmetry is described by an ansatz of the form

$$f(\psi) = A_0 \cos \psi + \delta.$$

1 a

Start with the ansatz

$$f(\psi) = a_1 f_1(\psi) + a_2 f_2(\psi) = a_1 \cos(\psi) + a_2 \sin(\psi)$$

and write down the design matrix \mathbf{A} .

$$\rightarrow \text{Designmatrix } \mathbf{A} = \begin{bmatrix} f_1(x_1) & \dots & f_n(x_1) \\ \vdots & & \vdots \\ f_1(x_n) & \dots & f_n(x_n) \end{bmatrix} \Rightarrow \text{hier: } \mathbf{A} = \begin{bmatrix} \cos(\psi_1) & \sin(\psi_1) \\ \vdots & \vdots \\ \cos(\psi_{42}) & \sin(\psi_{42}) \end{bmatrix} \text{ mit den Messwerten } \psi_1, \dots, \psi_{42} \text{ aus der Tabelle oben.}$$

```
import numpy as np
```

```
psi = np.radians(np.arange(0, 331, 30)) # muss damit numpy nicht weint in rad umgewan
measurement = np.array([-0.032, 0.010, 0.057, 0.068, 0.076, 0.080, 0.031, 0.005, -
0.041, -0.090, -0.088, -0.074])
measurement_error = 0.011
```

```
A = np.column_stack([np.cos(psi), np.sin(psi)])
print(f"A = {A}")
```

2 b

Calculate the solution vector \hat{a} for the parameters using the method of the least squares.

The solution vector \hat{a} can be calculated as $\hat{a} = (\mathbf{A}^T \cdot \mathbf{A})^{-1} \mathbf{A}^T \mathbf{y}$:

```
a = np.linalg.inv(A.T @ A) @ A.T @ measurement
print(f"a = {a}")
```

3 c

Calculate the covariance matrix $\mathbf{V}[\hat{\mathbf{a}}]$ as well as the errors of a_1 and a_2 and the correlation coefficient.

The covariance matrix of \hat{a} is given by $\mathbf{V}[\hat{\mathbf{a}}] = \sigma^2(\mathbf{A}^T \cdot \mathbf{A})^{-1}$, where σ denotes the measurement error.

```
V = measurement_error**2 * np.linalg.inv (A.T @ A)
print(f"V = {V}")
```

The error of a_1 and a_2 can be calculated via the diagonal elements of the covariance matrix:

```
a_error = np.sqrt(np.diag(V))
print(f"a_error = {a_error}")
```

4 d

Calculate A_0 and δ , their error and the correlation of a_1 and a_2 .

$$\begin{aligned}
 \rightarrow f(\psi) &= A_0 \cos(\psi + \delta) = a_1 \cos(\psi) + a_2 \sin(\psi) & \rightarrow \cos(x+y) &= \cos(x)\cos(y) - \sin(x)\sin(y) \\
 \Rightarrow A_0 \cos(\delta) \cos(\psi) - A_0 \sin(\delta) \sin(\psi) &= a_1 \cos(\psi) + a_2 \sin(\psi) \\
 \Rightarrow A_0 \cos(\delta) &= a_1, \quad -A_0 \sin(\delta) &= a_2 \\
 \downarrow \\
 A_0 &= \frac{a_1}{\cos(\delta)} \Rightarrow \frac{-a_1 \sin(\delta)}{\cos(\delta)} = a_2 \quad \Rightarrow \frac{\sin(\delta)}{\cos(\delta)} = \frac{-a_2}{a_1} \quad \Rightarrow \delta = \arctan\left(\frac{-a_2}{a_1}\right) \\
 \Rightarrow A_0 &= \frac{a_1}{\cos(\arctan(-\frac{a_2}{a_1}))} \quad \text{-- verwende } \cos(\tan^{-1}(\frac{-a}{b})) = \frac{1}{\sqrt{\frac{a^2}{b^2} + 1}} \\
 \Rightarrow A_0 &= a_1 \sqrt{\frac{a_1^2}{a_1^2} + 1}
 \end{aligned}$$

```
A_0 = a[0] * np.sqrt((a[1]**2 / a[0]**2) + 1)
delta = np.arctan(-a[1]/a[0])
print(f"A_0 = {A_0}")
print(f"delta = {delta}")
```

The error of a solution vector $\vec{y} = f(\vec{x})$ is can be computed by using $\mathbf{V}[\tilde{\mathbf{y}}] = \mathbf{J} \cdot \mathbf{V}[\tilde{\mathbf{x}}] \cdot \mathbf{J}^T$. Calculate the Jacobian matrix of $f(\vec{x}) = \begin{bmatrix} A_0 \\ \delta \end{bmatrix}$:

$$f(\vec{x}) = \begin{bmatrix} A_0 \\ \delta \end{bmatrix} = \begin{bmatrix} a_1 \sqrt{\frac{a_1^2}{a_1^2} - 1} \\ \arctan\left(\frac{-a_2}{a_1}\right) \end{bmatrix} = \begin{bmatrix} \sqrt{a_1^2 - a_1^2} \\ \arctan\left(\frac{-a_2}{a_1}\right) \end{bmatrix} \quad \rightarrow \text{Jacobian-Matrix } \mathbf{J} = \begin{bmatrix} \frac{\partial f_1}{\partial a_1} & \frac{\partial f_1}{\partial a_2} \\ \frac{\partial f_2}{\partial a_1} & \frac{\partial f_2}{\partial a_2} \end{bmatrix}$$

$$\frac{\partial f_1}{\partial a_1} = \frac{\partial}{\partial a_1} \sqrt{a_1^2 - a_1^2} = -\frac{a_1}{\sqrt{a_1^2 - a_1^2}} \quad , \quad \frac{\partial f_1}{\partial a_2} = \frac{\partial}{\partial a_2} \sqrt{a_1^2 - a_1^2} = \frac{a_2}{\sqrt{a_1^2 - a_1^2}}$$

$$\frac{\partial f_2}{\partial a_1} = \frac{\partial}{\partial a_1} \arctan\left(\frac{-a_2}{a_1}\right) = \frac{a_2}{a_1^2 + a_2^2} \quad , \quad \frac{\partial f_2}{\partial a_2} = \frac{\partial}{\partial a_2} \arctan\left(\frac{-a_2}{a_1}\right) = \frac{-a_1}{a_1^2 + a_2^2}$$

```
J = np.array([[(-a[0]/(np.sqrt(a[1]**2 - a[0]**2))), (a[1]/(np.sqrt(a[1]**2 - a[0]**2))
              [(a[1]/(a[0]**2 + a[1]**2)), (-a[0]/(a[0]**2 + a[1]**2))]])

V_2 = J @ V @ J.T
errors = np.sqrt(np.diag(V_2))
# correlation_coefficient = V_2[1,0] / V_2[0,0]
correlation_coefficient = V_2[1,0] / errors[0]**2

print(f"Fehler von A_0: {errors[0]}")
print(f"Fehler von delta: {errors[1]}")
print(f"Korrelationskoeffizient:{correlation_coefficient}")
```