Examples In [1]: import numpy as np import pandas as pd import scipy.stats as sts import matplotlib.pyplot as plt import seaborn as sns sns.set(style="whitegrid", font_scale=1.5) sns.despine() %matplotlib inline In [2]: def get_ax(): fig, ax = plt.subplots(figsize=(20,10)) return ax Example 1 In $[3]: N = 1_000_000$ theta = 0 population = np.random.randn(N) + theta get_ax().hist(population, bins=100); 40000 30000 20000 10000 0 6 -4 In $[4]: n = 1_{000}$ sample = np.random.choice(population, size=n, replace=False) estimate = sample.mean() print(f"mean-estimate: {estimate:5f}") mean-estimate: -0.004028 In [5]: $m = 1_000$ samples = np.array([np.random.choice(population, size=n, replace=False) for _ in range(m)]) sampling_distribution = estimates = np.array([sample.mean() for sample in samples]) get_ax().hist(sampling_distribution, bins=100); 40 35 30 25 20 15 10 5 0.00 -0.050.05 0.10 -0.10 In [6]: mean_estimate = sampling_distribution.mean() numerical_bias = mean_estimate - theta expected_estimate = theta analytical_bias = expected_estimate - theta print(f"Estimator bias (numerical): {numerical_bias:5f}") print(f"Estimator bias (analytical): {analytical_bias:5f}") Estimator bias (numerical): -0.000180 Estimator bias (analytical): 0.000000 In [7]: sem_numerical = np.sqrt(sampling_distribution.var()) std = sample.std() sem_numerical_std = std / np.sqrt(n) sem_scipy = sts.sem(sample) sem_analytical = np.sqrt(1 / n) print(f"Estimator standard error (numerical): {sem_numerical:5f}") print(f"Estimator standard error (numerical via std): {sem_numerical_std:5f}") print(f"Estimator standard error (numerical via scipy): {sem_scipy:5f}") print(f"Estimator standard error (analytical): {sem_analytical:5f}") Estimator standard error (numerical): 0.031657 Estimator standard error (numerical via std): 0.031115 Estimator standard error (numerical via scipy): 0.031131 Estimator standard error (analytical): 0.031623 Example 2 In [8]: $N = 1_{000}000$ theta = 1 population = np.random.rand(N) * theta get_ax().hist(population, bins=100); 10000 8000 6000 4000 2000 0 0.0 0.2 0.4 8.0 1.0 In $[9]: n = 1_{000}$ sample = np.random.choice(population, size=n, replace=False) estimate = sample.max() print(f"max-estimate: {estimate:5f}") max-estimate: 0.999887 In [10]: $m = 1_000$ samples = np.array([np.random.choice(population, size=n, replace=False) for _ in range(m)]) sampling_distribution = estimates = np.array([sample.max() for sample in samples]) get_ax().hist(sampling_distribution, bins=100); 80 60 40 20 0 0.997 0.998 0.999 0.993 0.994 0.995 0.996 1.000 In [11]: mean_estimate = sampling_distribution.mean() numerical_bias = mean_estimate - theta expected_estimate = n / (n + 1) * theta ** (n + 1)analytical_bias = expected_estimate - theta print(f"Estimator bias (numerical): {numerical_bias:5f}") print(f"Estimator bias (analytical): {analytical_bias:5f}") Estimator bias (numerical): -0.000973 Estimator bias (analytical): -0.000999 In [12]: sem_numerical = np.sqrt(sampling_distribution.var()) sem_analytical = np.sqrt(n / (n + 2) * theta ** 2 - expected_estimate ** 2) print(f"Estimator standard error (numerical): {sem_numerical:5f}") print(f"Estimator standard error (analytical): {sem_analytical:5f}") Estimator standard error (numerical): 0.000992 Estimator standard error (analytical): 0.000998 Can not use std. Why? std is for estimates of mean only. Example 3 In [13]: **import** sklearn.datasets **as** ds In [14]: wine_ds = ds.load_wine() df = pd.DataFrame(wine_ds["data"], columns=wine_ds["feature_names"]) df.head() Out[14]: alcohol malic_acid ash alcalinity_of_ash magnesium total_phenols flavanoids nonflavanoid_phenols proanthocyanins color_intensity hue od280/od315_of_diluted_wine 1.71 2.43 14.23 15.6 127.0 2.80 3.06 0.28 2.29 5.64 1.04 3.9 0 13.20 1.78 2.14 100.0 2.65 2.76 0.26 4.38 1.05 11.2 1.28 3.4 13.16 2.36 2.67 18.6 101.0 2.80 3.24 0.30 2.81 5.68 1.03 3.1 1.95 2.50 14.37 113.0 3.85 3.49 0.24 2.18 7.80 0.86 3.4 16.8 4 13.24 2.59 2.87 21.0 118.0 2.80 2.69 0.39 1.82 4.32 1.04 2.9 sample = df["magnesium"].transform(np.log).to_numpy() In [15]: get_ax().hist(sample, bins=100); 12 10 8 6 4.4 5.0 In [16]: def linear_statistical_functional(func): def lsf(x): return np.mean(func(x)) return 1sf In [17]: sample_mean = sample.mean() plug_in_estimator_for_mean = linear_statistical_functional(lambda x: x) print(f"Sample mean: {sample_mean:5f}") print(f"Plug-in estimator for mean: {plug_in_estimator_for_mean(sample):5f}") Sample mean: 4.593042 Plug-in estimator for mean: 4.593042 In [18]: sample_var = sample.var() plug_in_estimator_for_var = linear_statistical_functional(lambda x: (x - sample_mean) ** 2) print(f"Sample std: {sample_var:5f}") print(f"Plug-in estimator for std: {plug_in_estimator_for_var(sample):5f}") Sample std: 0.018562 Plug-in estimator for std: 0.018562