

# lognorm\_stat\_testi

August 21, 2021

```
[1]: import numpy as np
import matplotlib.pyplot as plt
from scipy.stats import lognorm

from benford_helper_functions import normalize, get_first_digit, benfords_test
```

```
[2]: p = 1
low, high = p * np.log(10), (p+1) * np.log(10)

x_ = np.random.uniform(low=low, high=high, size=100000)
x = np.exp(x_)

np.unique(get_first_digit(x), return_counts=True)
```

```
[2]: (array([1, 2, 3, 4, 5, 6, 7, 8, 9]),
      array([30284, 17461, 12395, 9715, 8027, 6693, 5708, 5108, 4609]))
```

```
[3]: from stat_tests import chi2_test, ks_test
```

```
[4]: import sys

sys.path.append('../..')
from plotting.matplotlib_setup import configure_latex, savefig,
    ↪set_size_decorator, savefig, thinner_border

tex_dir, images_dir = 'porocilo/main.tex', 'porocilo/images'

configure_latex(style=['science', 'notebook'], global_save_path=images_dir)

%config InlineBackend.figure_format = 'pdf'
```

```
[5]: def get_lognorm_pdf(x, mu, sigma):
    return (1 / (x * sigma * np.sqrt(2 * np.pi))) * np.exp(-((np.log10(x) -
    ↪mu)**2) / (2*sigma**2))

def lognormMu(x, mu, s):
    tempX = x / np.exp(mu)
```

```
return lognorm.pdf(tempX, s)
```

```
[6]: from scipy.stats import lognorm
```

```
def lognorm_params(mode, stddev):
```

```
    """
```

*Given the mode and std. dev. of the log-normal distribution, this function returns the shape and scale parameters for scipy's parameterization of the distribution.*

*<https://stackoverflow.com/questions/41464753/>*

*→ generate-random-numbers-from-lognormal-distribution-in-python*

```
    """
```

```
    p = np.poly1d([1, -1, 0, 0, -(stddev/mode)**2])
```

```
    r = p.roots
```

```
    sol = r[(r.imag == 0) & (r.real > 0)].real
```

```
    shape = np.sqrt(np.log(sol))
```

```
    scale = mode * sol
```

```
    return shape, scale
```

```
def lognorm_params_exact(mode, stddev):
```

```
    a = stddev**2 / mode**2
```

```
    x = 1/4*np.sqrt(-(16*(2/3)**(1/3)*a)/(np.sqrt(3)*np.
```

```
    → sqrt(256*a**3+27*a**2)-9*a)**(1/3) +
```

```
        2*(2/3)**(2/3)*(np.sqrt(3)*np.
```

```
    → sqrt(256*a**3+27*a**2)-9*a)**(1/3)+1) + \
```

```
        1/2*np.sqrt((4*(2/3)**(1/3)*a)/(np.sqrt(3)*np.
```

```
    → sqrt(256*a**3+27*a**2)-9*a)**(1/3) -
```

```
        (np.sqrt(3)*np.sqrt(256*a**3+27*a**2)-9*a)**(1/3)/(2**((1/
```

```
    → 3)*3**((2/3)) +
```

```
        1/(2*np.sqrt(-(16*(2/3)**(1/3)*a)/(np.sqrt(3)*np.
```

```
    → sqrt(256*a**3+27*a**2)-9*a)**(1/3) +
```

```
        2*(2/3)**(2/3)*(np.sqrt(3)*np.
```

```
    → sqrt(256*a**3+27*a**2)-9*a)**(1/3)+1))+1/2) + \
```

```
        1/4
```

```
    shape = np.sqrt(np.log(x))
```

```
    scale = mode * x
```

```
    return shape, scale
```

```
N = 10**6
```

```
mu = 10**3
```

```
# sigmas = np.array([100, 1000, 10000, 100000]) # -> [0.5, 0.8, 1.2]
```

```

sigmas = np.logspace(2, 12, 100)

lognorm_dists, lognorm_sigmas = [], []
for s in sigmas:
    sigma, scale = lognorm_params(mu, s)
    lognorm_sigmas.append(sigma[0])

    np.random.seed(1)
    lognorm_rng = lognorm.rvs(sigma, 0, scale, size=N)

    lognorm_dists.append(lognorm_rng)

```

```

[7]: f1s = []

for i, lognorm in enumerate(lognorm_dists):
    bins = np.logspace(np.floor(np.log10(lognorm.min())) ,
                        np.floor(np.log10(lognorm.max())) + 1,
                        len(lognorm))

    pdf, _ = np.histogram(lognorm, bins=bins)

    bins = bins[:-1]
    bins = np.log10(bins)
    pdf = normalize(pdf, bins)

    f1 = benford_test(pdf, bins)
    f1s.append(f1)

    first_digits = get_first_digit(lognorm)
    _, n = np.unique(first_digits, return_counts=True)
    n = n / np.sum(n)

    print(f'delez 1: {n[0]:.4f}, f1: {f1:.3e}, normal sigma: {sigmas[i]:.2f},
    ↪ lognorm sigma: {lognorm_sigmas[i]:.2f}')

```

```

delez 1: 0.5393, f1: 9.647e-01, normal sigma: 100.00, lognorm sigma: 0.10
delez 1: 0.5489, f1: 9.454e-01, normal sigma: 126.19, lognorm sigma: 0.12
delez 1: 0.5607, f1: 9.168e-01, normal sigma: 159.23, lognorm sigma: 0.15
delez 1: 0.5747, f1: 8.759e-01, normal sigma: 200.92, lognorm sigma: 0.19
delez 1: 0.5888, f1: 8.201e-01, normal sigma: 253.54, lognorm sigma: 0.23
delez 1: 0.5964, f1: 7.484e-01, normal sigma: 319.93, lognorm sigma: 0.28
delez 1: 0.5904, f1: 6.627e-01, normal sigma: 403.70, lognorm sigma: 0.33
delez 1: 0.5688, f1: 5.676e-01, normal sigma: 509.41, lognorm sigma: 0.39
delez 1: 0.5355, f1: 4.701e-01, normal sigma: 642.81, lognorm sigma: 0.45
delez 1: 0.4959, f1: 3.769e-01, normal sigma: 811.13, lognorm sigma: 0.51
delez 1: 0.4541, f1: 2.934e-01, normal sigma: 1023.53, lognorm sigma: 0.57
delez 1: 0.4150, f1: 2.224e-01, normal sigma: 1291.55, lognorm sigma: 0.64

```

delez 1: 0.3801, f1: 1.648e-01, normal sigma: 1629.75, lognorm sigma: 0.70  
delez 1: 0.3517, f1: 1.197e-01, normal sigma: 2056.51, lognorm sigma: 0.75  
delez 1: 0.3307, f1: 8.552e-02, normal sigma: 2595.02, lognorm sigma: 0.81  
delez 1: 0.3154, f1: 6.018e-02, normal sigma: 3274.55, lognorm sigma: 0.87  
delez 1: 0.3057, f1: 4.179e-02, normal sigma: 4132.01, lognorm sigma: 0.92  
delez 1: 0.2998, f1: 2.867e-02, normal sigma: 5214.01, lognorm sigma: 0.97  
delez 1: 0.2977, f1: 1.943e-02, normal sigma: 6579.33, lognorm sigma: 1.02  
delez 1: 0.2973, f1: 1.299e-02, normal sigma: 8302.18, lognorm sigma: 1.07  
delez 1: 0.2978, f1: 8.545e-03, normal sigma: 10476.16, lognorm sigma: 1.12  
delez 1: 0.2984, f1: 5.491e-03, normal sigma: 13219.41, lognorm sigma: 1.17  
delez 1: 0.2992, f1: 3.401e-03, normal sigma: 16681.01, lognorm sigma: 1.21  
delez 1: 0.2995, f1: 1.975e-03, normal sigma: 21049.04, lognorm sigma: 1.26  
delez 1: 0.3001, f1: 1.009e-03, normal sigma: 26560.88, lognorm sigma: 1.30  
delez 1: 0.3008, f1: 3.902e-04, normal sigma: 33516.03, lognorm sigma: 1.34  
delez 1: 0.3017, f1: 2.684e-04, normal sigma: 42292.43, lognorm sigma: 1.38  
delez 1: 0.3015, f1: 4.985e-04, normal sigma: 53366.99, lognorm sigma: 1.42  
delez 1: 0.3014, f1: 6.808e-04, normal sigma: 67341.51, lognorm sigma: 1.46  
delez 1: 0.3010, f1: 7.976e-04, normal sigma: 84975.34, lognorm sigma: 1.50  
delez 1: 0.3011, f1: 8.630e-04, normal sigma: 107226.72, lognorm sigma: 1.54  
delez 1: 0.3004, f1: 8.899e-04, normal sigma: 135304.78, lognorm sigma: 1.57  
delez 1: 0.3004, f1: 8.887e-04, normal sigma: 170735.26, lognorm sigma: 1.61  
delez 1: 0.3010, f1: 8.670e-04, normal sigma: 215443.47, lognorm sigma: 1.64  
delez 1: 0.3009, f1: 8.311e-04, normal sigma: 271858.82, lognorm sigma: 1.68  
delez 1: 0.3010, f1: 7.858e-04, normal sigma: 343046.93, lognorm sigma: 1.71  
delez 1: 0.3009, f1: 7.351e-04, normal sigma: 432876.13, lognorm sigma: 1.75  
delez 1: 0.3007, f1: 6.821e-04, normal sigma: 546227.72, lognorm sigma: 1.78  
delez 1: 0.3008, f1: 6.293e-04, normal sigma: 689261.21, lognorm sigma: 1.81  
delez 1: 0.3002, f1: 5.790e-04, normal sigma: 869749.00, lognorm sigma: 1.84  
delez 1: 0.3008, f1: 5.328e-04, normal sigma: 1097498.77, lognorm sigma: 1.87  
delez 1: 0.3012, f1: 4.922e-04, normal sigma: 1384886.37, lognorm sigma: 1.90  
delez 1: 0.3020, f1: 4.586e-04, normal sigma: 1747528.40, lognorm sigma: 1.93  
delez 1: 0.3017, f1: 4.329e-04, normal sigma: 2205130.74, lognorm sigma: 1.96  
delez 1: 0.3015, f1: 4.158e-04, normal sigma: 2782559.40, lognorm sigma: 1.99  
delez 1: 0.3014, f1: 4.073e-04, normal sigma: 3511191.73, lognorm sigma: 2.02  
delez 1: 0.3010, f1: 4.074e-04, normal sigma: 4430621.46, lognorm sigma: 2.05  
delez 1: 0.3007, f1: 4.155e-04, normal sigma: 5590810.18, lognorm sigma: 2.08  
delez 1: 0.3007, f1: 4.308e-04, normal sigma: 7054802.31, lognorm sigma: 2.11  
delez 1: 0.3005, f1: 4.529e-04, normal sigma: 8902150.85, lognorm sigma: 2.13  
delez 1: 0.3010, f1: 4.812e-04, normal sigma: 11233240.33, lognorm sigma: 2.16  
delez 1: 0.3008, f1: 5.156e-04, normal sigma: 14174741.63, lognorm sigma: 2.19  
delez 1: 0.3008, f1: 5.557e-04, normal sigma: 17886495.29, lognorm sigma: 2.21  
delez 1: 0.3006, f1: 6.015e-04, normal sigma: 22570197.20, lognorm sigma: 2.24  
delez 1: 0.3008, f1: 6.527e-04, normal sigma: 28480358.68, lognorm sigma: 2.26  
delez 1: 0.3011, f1: 7.087e-04, normal sigma: 35938136.64, lognorm sigma: 2.29  
delez 1: 0.3010, f1: 7.688e-04, normal sigma: 45348785.08, lognorm sigma: 2.32  
delez 1: 0.3009, f1: 8.320e-04, normal sigma: 57223676.59, lognorm sigma: 2.34  
delez 1: 0.3013, f1: 8.969e-04, normal sigma: 72208090.18, lognorm sigma: 2.37  
delez 1: 0.3015, f1: 9.623e-04, normal sigma: 91116275.61, lognorm sigma: 2.39

delez 1: 0.3007, f1: 1.026e-03, normal sigma: 114975699.54, lognorm sigma: 2.41  
delez 1: 0.3009, f1: 1.088e-03, normal sigma: 145082877.85, lognorm sigma: 2.44  
delez 1: 0.3012, f1: 1.145e-03, normal sigma: 183073828.03, lognorm sigma: 2.46  
delez 1: 0.3018, f1: 1.196e-03, normal sigma: 231012970.01, lognorm sigma: 2.49  
delez 1: 0.3021, f1: 1.240e-03, normal sigma: 291505306.28, lognorm sigma: 2.51  
delez 1: 0.3021, f1: 1.275e-03, normal sigma: 367837977.18, lognorm sigma: 2.53  
delez 1: 0.3021, f1: 1.300e-03, normal sigma: 464158883.36, lognorm sigma: 2.55  
delez 1: 0.3013, f1: 1.315e-03, normal sigma: 585702081.81, lognorm sigma: 2.58  
delez 1: 0.3013, f1: 1.319e-03, normal sigma: 739072203.35, lognorm sigma: 2.60  
delez 1: 0.3001, f1: 1.312e-03, normal sigma: 932603346.88, lognorm sigma: 2.62  
delez 1: 0.3003, f1: 1.293e-03, normal sigma: 1176811952.43, lognorm sigma: 2.64  
delez 1: 0.3001, f1: 1.263e-03, normal sigma: 1484968262.25, lognorm sigma: 2.67  
delez 1: 0.3001, f1: 1.222e-03, normal sigma: 1873817422.86, lognorm sigma: 2.69  
delez 1: 0.3003, f1: 1.171e-03, normal sigma: 2364489412.65, lognorm sigma: 2.71  
delez 1: 0.3011, f1: 1.112e-03, normal sigma: 2983647240.28, lognorm sigma: 2.73  
delez 1: 0.3010, f1: 1.044e-03, normal sigma: 3764935806.79, lognorm sigma: 2.75  
delez 1: 0.3011, f1: 9.685e-04, normal sigma: 4750810162.10, lognorm sigma: 2.77  
delez 1: 0.3007, f1: 8.877e-04, normal sigma: 5994842503.19, lognorm sigma: 2.79  
delez 1: 0.3010, f1: 8.025e-04, normal sigma: 7564633275.55, lognorm sigma: 2.81  
delez 1: 0.3010, f1: 7.142e-04, normal sigma: 9545484566.62, lognorm sigma: 2.83  
delez 1: 0.3012, f1: 6.246e-04, normal sigma: 12045035402.59, lognorm sigma:  
2.86  
delez 1: 0.3013, f1: 5.351e-04, normal sigma: 15199110829.53, lognorm sigma:  
2.88  
delez 1: 0.3012, f1: 4.477e-04, normal sigma: 19179102616.72, lognorm sigma:  
2.90  
delez 1: 0.3013, f1: 3.650e-04, normal sigma: 24201282647.94, lognorm sigma:  
2.92  
delez 1: 0.3013, f1: 2.909e-04, normal sigma: 30538555088.33, lognorm sigma:  
2.94  
delez 1: 0.3012, f1: 2.323e-04, normal sigma: 38535285937.11, lognorm sigma:  
2.96  
delez 1: 0.3009, f1: 1.997e-04, normal sigma: 48626015800.65, lognorm sigma:  
2.97  
delez 1: 0.3007, f1: 2.018e-04, normal sigma: 61359072734.13, lognorm sigma:  
2.99  
delez 1: 0.3004, f1: 2.336e-04, normal sigma: 77426368268.11, lognorm sigma:  
3.01  
delez 1: 0.3009, f1: 2.817e-04, normal sigma: 97700995729.92, lognorm sigma:  
3.03  
delez 1: 0.3015, f1: 3.361e-04, normal sigma: 123284673944.21, lognorm sigma:  
3.05  
delez 1: 0.3016, f1: 3.919e-04, normal sigma: 155567614393.05, lognorm sigma:  
3.07  
delez 1: 0.3013, f1: 4.464e-04, normal sigma: 196304065004.03, lognorm sigma:  
3.09  
delez 1: 0.3010, f1: 4.985e-04, normal sigma: 247707635599.17, lognorm sigma:  
3.11

```

delez 1: 0.3015, f1: 5.476e-04, normal sigma: 312571584968.82, lognorm sigma:
3.13
delez 1: 0.3013, f1: 5.935e-04, normal sigma: 394420605943.76, lognorm sigma:
3.15
delez 1: 0.3010, f1: 6.363e-04, normal sigma: 497702356433.21, lognorm sigma:
3.16
delez 1: 0.3014, f1: 6.760e-04, normal sigma: 628029144183.42, lognorm sigma:
3.18
delez 1: 0.3009, f1: 7.130e-04, normal sigma: 792482898353.92, lognorm sigma:
3.20
delez 1: 0.3011, f1: 7.472e-04, normal sigma: 1000000000000.00, lognorm sigma:
3.22

```

(Uniform distribution characterization). A sequence of real numbers (respectively, a Borel measurable function, a random variable, a Borel probability measure) is Benford if and only if the decimal logarithm of its absolute value is uniformly distributed modulo 1.

```
[8]: sigmas = np.sqrt(np.log10(1 + sigmas / mu**2))
```

```
[9]: lognorm_fracs = []

for dist in lognorm_dists:
    # lognorm_fracs.append(get_number_fracs_math(np.log10(dist[dist > 0])))
    lognorm_fracs.append(np.log10(dist) % 1)
```

```
[10]: lognorm_fracs_hists = []

for i, fracs in enumerate(lognorm_fracs):
    n, bins, _ = plt.hist(fracs, bins=30, density=False, histtype='step')
    lognorm_fracs_hists.append([n, bins[1:]])

plt.close()
```

## 1 Test $\chi^2$

```
[11]: # chi2_test(np.array(lognorm_fracs), n_bins=30)
```

```
[12]: from scipy.stats import chisquare
from scipy.stats import chi2
```

```
[13]: chi2_test = []

N, n_bins = np.sum(lognorm_fracs_hists[0][0]), len(lognorm_fracs_hists[0][1])

for hist in lognorm_fracs_hists:
    chi2_ = chisquare(hist[0], f_exp=N/n_bins)
    chi2_test.append([chi2_.statistic, chi2_.pvalue])
```

```
[14]: for c in chi2_test:
      print(f'{c[0]:.3f}, {c[1]:.3f}')
```

```
5445502.694, 0.000
4201962.289, 0.000
3207930.557, 0.000
2417048.101, 0.000
1800594.890, 0.000
1320276.298, 0.000
948841.878, 0.000
663092.915, 0.000
445008.712, 0.000
283916.566, 0.000
171555.056, 0.000
98570.562, 0.000
54121.416, 0.000
28610.910, 0.000
14648.218, 0.000
7242.391, 0.000
3508.612, 0.000
1675.201, 0.000
794.073, 0.000
381.696, 0.000
174.293, 0.000
95.251, 0.000
56.577, 0.002
49.820, 0.009
32.397, 0.303
52.016, 0.005
34.217, 0.231
28.004, 0.518
42.637, 0.049
25.386, 0.658
28.536, 0.489
32.745, 0.288
31.983, 0.321
37.829, 0.126
47.098, 0.018
41.112, 0.067
32.430, 0.301
34.468, 0.223
21.481, 0.841
46.234, 0.022
40.160, 0.081
49.189, 0.011
37.330, 0.138
27.647, 0.537
```

33.843, 0.245  
29.176, 0.456  
31.339, 0.350  
27.141, 0.564  
40.260, 0.080  
26.219, 0.614  
25.656, 0.644  
38.929, 0.103  
25.860, 0.633  
27.570, 0.541  
32.804, 0.286  
34.197, 0.232  
51.092, 0.007  
35.972, 0.174  
25.474, 0.653  
36.920, 0.148  
42.216, 0.054  
38.056, 0.121  
35.618, 0.185  
28.619, 0.485  
45.330, 0.027  
39.287, 0.096  
52.800, 0.004  
44.134, 0.036  
45.431, 0.027  
46.371, 0.022  
47.700, 0.016  
57.100, 0.001  
39.289, 0.096  
39.671, 0.089  
30.650, 0.382  
30.227, 0.403  
27.182, 0.562  
36.655, 0.155  
34.635, 0.217  
32.177, 0.312  
25.302, 0.662  
34.494, 0.222  
34.967, 0.206  
32.174, 0.312  
30.365, 0.396  
33.923, 0.242  
36.046, 0.172  
19.678, 0.903  
19.631, 0.904  
29.615, 0.433  
26.657, 0.590  
33.142, 0.272



```

37.524, 0.133
50.174, 0.009
49.563, 0.010
27.185, 0.562
29.716, 0.428
24.375, 0.710
46.138, 0.023
30.857, 0.372

```

```

[15]: alpha = 0.01 # stopnja pomembnosti

# stopnja zaupanja
critical_value_chi2 = chi2.ppf(1 - alpha, n_bins)
critical_value_chi2

```

```

[15]: 50.89218131151707

```

## 2 Test Kolmogorov-Smirnova

```

[16]: ks_test(np.array(lognorm_fracs))

```

```

[16]: (array([[4.27647627e-001, 0.00000000e+000],
 [4.12863214e-001, 0.00000000e+000],
 [3.95750415e-001, 0.00000000e+000],
 [3.76083786e-001, 0.00000000e+000],
 [3.53442606e-001, 0.00000000e+000],
 [3.27730384e-001, 0.00000000e+000],
 [2.99392307e-001, 0.00000000e+000],
 [2.68281036e-001, 0.00000000e+000],
 [2.36003538e-001, 0.00000000e+000],
 [2.01724347e-001, 0.00000000e+000],
 [1.67375939e-001, 0.00000000e+000],
 [1.34319390e-001, 0.00000000e+000],
 [1.03688112e-001, 0.00000000e+000],
 [7.66349102e-002, 0.00000000e+000],
 [5.43322610e-002, 0.00000000e+000],
 [3.66844702e-002, 0.00000000e+000],
 [2.34325619e-002, 0.00000000e+000],
 [1.38814678e-002, 8.26037701e-168],
 [7.85264784e-003, 5.46130671e-054],
 [4.36224377e-003, 5.90436355e-017],
 [3.41470867e-003, 1.48616029e-010],
 [3.01190645e-003, 2.63432380e-008],
 [2.13730047e-003, 2.15103434e-004],
 [1.70640438e-003, 5.90695544e-003],
 [1.28047034e-003, 7.52477854e-002],

```

[7.53866941e-004, 6.20380797e-001],  
[9.64710748e-004, 3.09565203e-001],  
[1.06724679e-003, 2.04605714e-001],  
[7.69549439e-004, 5.94102102e-001],  
[5.62185952e-004, 9.09877341e-001],  
[7.83946809e-004, 5.70193011e-001],  
[1.02269533e-003, 2.46294833e-001],  
[9.13742392e-004, 3.73814857e-001],  
[1.03190945e-003, 2.37190224e-001],  
[7.79679821e-004, 5.77252644e-001],  
[1.06901042e-003, 2.03074412e-001],  
[9.29362921e-004, 3.53269993e-001],  
[1.28699522e-003, 7.27684466e-002],  
[8.51206331e-004, 4.63235828e-001],  
[8.85191585e-004, 4.13261174e-001],  
[8.21507453e-004, 5.09314894e-001],  
[7.45528631e-004, 6.34417805e-001],  
[9.67592596e-004, 3.06173914e-001],  
[1.14535074e-003, 1.44908311e-001],  
[9.15331272e-004, 3.71691085e-001],  
[7.98210018e-004, 5.46785939e-001],  
[5.94710852e-004, 8.70992100e-001],  
[7.80684613e-004, 5.75588090e-001],  
[7.30195677e-004, 6.60275523e-001],  
[7.13839986e-004, 6.87809744e-001],  
[5.94551501e-004, 8.71198420e-001],  
[5.23498924e-004, 9.46763613e-001],  
[7.10020619e-004, 6.94216426e-001],  
[8.77858425e-004, 4.23775796e-001],  
[8.89463835e-004, 4.07206267e-001],  
[8.33730821e-004, 4.90100696e-001],  
[8.73992182e-004, 4.29380390e-001],  
[1.03989612e-003, 2.29504396e-001],  
[8.32180114e-004, 4.92519745e-001],  
[6.84492962e-004, 7.36597923e-001],  
[8.28253183e-004, 4.98670094e-001],  
[8.18047960e-004, 5.14811723e-001],  
[7.75976417e-004, 5.83398629e-001],  
[1.00383416e-003, 2.65735687e-001],  
[1.10320387e-003, 1.75103396e-001],  
[1.33330723e-003, 5.70868257e-002],  
[1.65095778e-003, 8.57178979e-003],  
[1.20934989e-003, 1.07220499e-001],  
[1.05602407e-003, 2.14558992e-001],  
[9.75213375e-004, 2.97331053e-001],  
[1.04774804e-003, 2.22132784e-001],  
[1.02813996e-003, 2.40884011e-001],



[illegible]

[illegible]

```
[17]: from scipy.stats import kstest
      from scipy.stats import kstwo
```

```
[18]: ks_test = []

for dist in lognorm_fracs:
    ks = kstest(dist, cdf='uniform', alternative='two-sided', args=(0, 1))
    stat, p = ks.statistic, ks.pvalue
    ks_test.append([stat, p])
    print(f'{stat:.2e}, {p:.3f}')
```

4.28e-01, 0.000

4.13e-01, 0.000  
3.96e-01, 0.000  
3.76e-01, 0.000  
3.53e-01, 0.000  
3.28e-01, 0.000  
2.99e-01, 0.000  
2.68e-01, 0.000  
2.36e-01, 0.000  
2.02e-01, 0.000  
1.67e-01, 0.000  
1.34e-01, 0.000  
1.04e-01, 0.000  
7.66e-02, 0.000  
5.43e-02, 0.000  
3.67e-02, 0.000  
2.34e-02, 0.000  
1.39e-02, 0.000  
7.85e-03, 0.000  
4.36e-03, 0.000  
3.41e-03, 0.000  
3.01e-03, 0.000  
2.14e-03, 0.000  
1.71e-03, 0.006  
1.28e-03, 0.075  
7.54e-04, 0.620  
9.65e-04, 0.310  
1.07e-03, 0.205  
7.70e-04, 0.594  
5.62e-04, 0.910  
7.84e-04, 0.570  
1.02e-03, 0.246  
9.14e-04, 0.374  
1.03e-03, 0.237  
7.80e-04, 0.577  
1.07e-03, 0.203  
9.29e-04, 0.353  
1.29e-03, 0.073  
8.51e-04, 0.463  
8.85e-04, 0.413  
8.22e-04, 0.509  
7.46e-04, 0.634  
9.68e-04, 0.306  
1.15e-03, 0.145  
9.15e-04, 0.372  
7.98e-04, 0.547  
5.95e-04, 0.871  
7.81e-04, 0.576  
7.30e-04, 0.660

7.14e-04, 0.688  
5.95e-04, 0.871  
5.23e-04, 0.947  
7.10e-04, 0.694  
8.78e-04, 0.424  
8.89e-04, 0.407  
8.34e-04, 0.490  
8.74e-04, 0.429  
1.04e-03, 0.230  
8.32e-04, 0.493  
6.84e-04, 0.737  
8.28e-04, 0.499  
8.18e-04, 0.515  
7.76e-04, 0.583  
1.00e-03, 0.266  
1.10e-03, 0.175  
1.33e-03, 0.057  
1.65e-03, 0.009  
1.21e-03, 0.107  
1.06e-03, 0.215  
9.75e-04, 0.297  
1.05e-03, 0.222  
1.03e-03, 0.241  
1.31e-03, 0.063  
1.29e-03, 0.072  
9.07e-04, 0.383  
9.36e-04, 0.344  
7.36e-04, 0.650  
9.78e-04, 0.294  
9.17e-04, 0.370  
9.07e-04, 0.383  
7.30e-04, 0.660  
6.57e-04, 0.781  
8.05e-04, 0.536  
7.28e-04, 0.664  
6.24e-04, 0.830  
5.61e-04, 0.912  
5.11e-04, 0.957  
6.09e-04, 0.852  
7.89e-04, 0.562  
5.68e-04, 0.903  
6.78e-04, 0.746  
8.14e-04, 0.521  
6.78e-04, 0.748  
8.42e-04, 0.478  
7.66e-04, 0.601  
8.66e-04, 0.441  
7.16e-04, 0.684

```
8.88e-04, 0.409
7.73e-04, 0.589
7.32e-04, 0.657
```

```
[19]: critical_value_ks = kstwo.ppf(1 - alpha, len(lognorm_fracs[0]))
      f'{critical_value_ks:.2e}'
```

```
[19]: '1.63e-03'
```

```
[20]: if False:
      fig, axs = set_size_decorator(plt.subplots, fraction=1, ratio='4:3')(2, 2)
      axs = axs.flatten()

      for i, fracs in enumerate(lognorm_fracs):
          axs[i].hist(fracs, bins=30, density=False, histtype='step')
          lognorm_fracs_hists.append([n, bins[1:]])

          # plt.close()

          axs[i].ticklabel_format(style='sci', axis='y', scilimits=(0, 0))

          an1 = f'$\sigma$ = {sigmas[i]:.2f}, $\sigma_X$ = {lognorm_sigmas[i]:.2f}'
          an2 = f'\n$\chi^2$ = {chi2_test[i][0]:.2f}, $d$ = {ks_test[i][0]:.2e}'
          an3 = f'\n$\chi^2_*$ = {critical_value_chi2:.2f}, $d_*$ = {critical_value_ks:.2e}'
          an4 = r' pri $\alpha$ = {}'.format(alpha)

          if i == 3:
              an = an1 + an2 + an3 + an4
          else:
              an = an1 + an2

          axs[i].annotate(an, xy=(0.1, 0.1), xycoords='axes fraction', fontsize=8)

      savefig('lognorm_uniform_hists')
```

```
[21]: fig, ax = set_size_decorator(plt.subplots, fraction=0.5, ratio='4:3')(1, 1)

      ax.set_yscale('log')

      x = lognorm_sigmas

      y = [i[0] for i in chi2_test]
      ax.scatter(x, y, s=5)
      ax.plot(x, y, lw=1)
```



```

ax.set_xlabel('$\sigma_X$')

ax.set_ylabel('$\chi^2$', c='C0')
ax.tick_params(axis='y', labelcolor='C0')

ax.set_ylim([0.9e1, 1e7])

ax2 = ax.twinx()

ax2.set_yscale('log')
ax2.minorticks_off()
ax2 = thinner_border(ax2)

y2 = [i[0] for i in ks_test]
ax2.scatter(x, y2, s=5, c='C2')
ax2.plot(x, y2, lw=1, c='C2')

ax2.set_ylabel('$d$', c='C2')
ax2.tick_params(axis='y', labelcolor='C2')

x3 = sigmas

ax3 = ax.twiny()
ax3 = thinner_border(ax3)
# ax3.set_xscale('log')

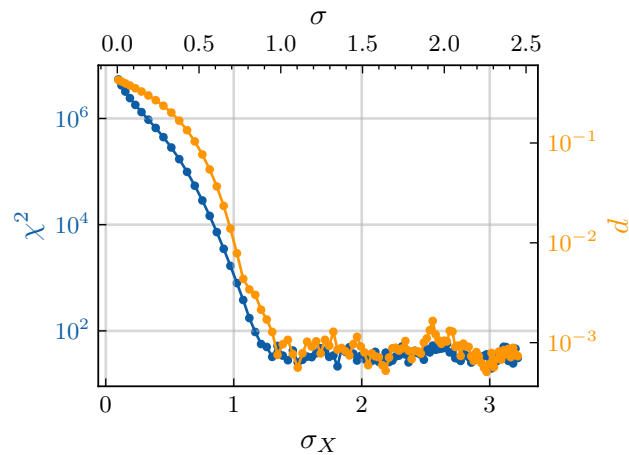
ax3.plot(x3, y, lw=0)

ax3.set_xlabel('$\sigma$')

ax.grid(zorder=0, alpha=0.5)

# savefig('stat_lognorm_tests', tight_layout=False)

```



```
[22]: chi2_lst = np.array([i[0] for i in chi2_test])
      ks_lst = np.array([i[0] for i in ks_test])
```

```
[23]: f1s = np.array(f1s)
```

```
[24]: idx = np.argsort(f1s)

fig, ax = set_size_decorator(plt.subplots, fraction=0.5, ratio='4:3')(1, 1)

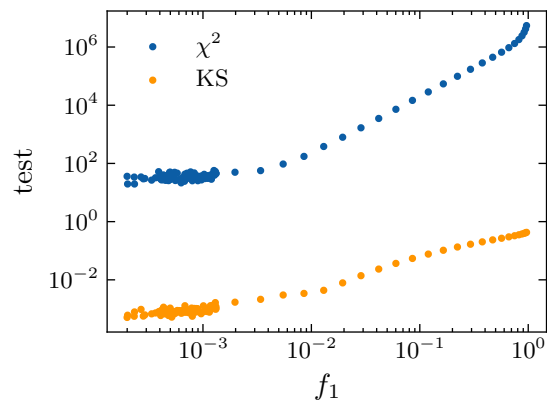
ax.set_xscale('log')
ax.set_yscale('log')
ax.scatter(f1s[idx], chi2_lst[idx], s=3, label='$\chi^2$')
ax.scatter(f1s[idx], ks_lst[idx], s=3, label='KS', c='C2')

ax.set_ylabel('test')
ax.set_xlabel('$f_1$')

ax.legend()

# savefig('test_lognorm_f1_chi2_KS')
```

```
[24]: <matplotlib.legend.Legend at 0x7f8d539fbc10>
```



```
[ ]:
```