cov-matrix-processing

July 1, 2022

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[1]: import numpy as np
     import matplotlib.pyplot as plt
     %matplotlib inline
[2]: cov_nercome = np.loadtxt("../output/Patchy_V6C_BOSS_DR12_NGC_z1_cov_nercome_200.
     ⇔matrix")
     cov_sample = np.loadtxt("../output/Patchy_V6C_BOSS_DR12_NGC_z1_cov_sample_200.
      →matrix")
     # 'Real' covariance matrix before processing.
     # The term 'real' here indicates that this is the matrix that we use to compare
     # our estimates to.
     cov real preproc = np.loadtxt("../data/
      -C_2048_BOSS_DR12_NGC_z1_V6C_1_1_1_1_1_10_200_200_prerecon.matrix")
[3]: print(cov_nercome)
     print(cov_sample)
    [[2.28506959e+08 6.42655535e+06 -1.06423240e+06 ... -4.30511096e+03]
      -6.29546771e+04 -3.67480992e+04]
     [ 6.42655535e+06 5.72291300e+07 1.89950886e+06 ... 2.83303162e+04
       7.40813638e+04 -1.25777799e+04]
     [-1.06423240e+06 1.89950886e+06 2.11413010e+07 ... -5.56466397e+03
      -2.42813916e+04 -2.76331372e+04]
     [-4.30511096e+03 2.83303162e+04 -5.56466397e+03 ... 1.13682080e+04
       2.89956568e+03 -6.22692178e+01]
     [-6.29546771e+04 7.40813638e+04 -2.42813916e+04 ... 2.89956568e+03
       1.06798482e+04 1.39264414e+03]
     [-3.67480992e+04 -1.25777799e+04 -2.76331372e+04 ... -6.22692178e+01
       1.39264414e+03 9.25917354e+03]]
    [[ 2.19870449e+08 6.58146060e+06 -8.49756132e+05 ... -6.25446109e+03
      -5.98302303e+04 -3.80366068e+04]
     [ 6.58146060e+06 5.14587823e+07 2.31434979e+06 ... 2.35449432e+04
       6.52292161e+04 -1.18367386e+04]
     [-8.49756132e+05 2.31434979e+06 1.78541839e+07 ... -4.12673501e+03
      -1.98163085e+04 -2.45021651e+04]
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[-6.25446109e+03 2.35449432e+04 -4.12673501e+03 ... 4.11567779e+03
       1.53786167e+03 2.02838640e-01]
     [-5.98302303e+04 6.52292161e+04 -1.98163085e+04 ... 1.53786167e+03
       3.71231389e+03 6.81792792e+02]
     [-3.80366068e+04 -1.18367386e+04 -2.45021651e+04 ... 2.02838640e-01
       6.81792792e+02 2.80471896e+03]]
[4]: indices = np.concatenate((np.arange(40), np.arange(40)+80, np.arange(40)+160))
    cov real = (cov real preproc[indices, :])[:, indices]
[5]: print(cov_real)
    -1.21684291e+04 2.21567909e+04]
     [ 1.91051563e+07   5.15256710e+07   4.78376836e+06 ...   9.13608077e+03
      9.26311893e+03 -1.41639754e+04]
     [ 1.44038639e+05 4.78376836e+06 1.67772970e+07 ... -5.49127765e+02
     -2.28200815e+03 6.76773412e+03]
     [ 9.68083967e+03  9.13608077e+03 -5.49127765e+02 ... 4.22979902e+03
       1.28931823e+03 3.78440184e+02]
     [-1.21684291e+04 9.26311893e+03 -2.28200815e+03 ... 1.28931823e+03
      3.88691287e+03 1.00320951e+03]
     [ 2.21567909e+04 -1.41639754e+04 6.76773412e+03 ... 3.78440184e+02
       1.00320951e+03 3.62984716e+03]]
[6]: evals_nercome = np.linalg.eigvalsh(cov_nercome)
    evals_sample = np.linalg.eigvalsh(cov_sample)
    evals_real = np.linalg.eigvalsh(cov_real)
[7]: plt.figure(dpi=200)
    plt.plot(evals_real, label="Beutler")
    plt.plot(evals_sample, label="Sample")
    plt.plot(evals_nercome, label="NERCOME")
    plt.yscale("log")
    plt.legend()
    plt.xlabel("Eigenmodes")
    plt.ylabel("Eigenvalues")
[7]: Text(0, 0.5, 'Eigenvalues')
```

```
10<sup>9</sup>
                       Beutler
                       Sample
    10<sup>8</sup>
                       NERCOME
    10^{7}
Eigenvalues
    10^{6}
   10<sup>5</sup>
    10^{4}
    10^{3}
    10^{2}
                              20
               0
                                              40
                                                              60
                                                                              80
                                                                                             100
                                                                                                             120
                                                      Eigenmodes
```

```
[8]: nercome_rel_errors = (cov_nercome-cov_real)/np.abs(cov_real)
     print(nercome_rel_errors)
    [[-0.04602701 -0.66362194 -8.38852026 ... -1.44470429 -4.17360757
      -2.65854791]
     0.11198802]
     [-8.38852026 -0.60292625 0.26011365 ... -9.13364161 -9.6403615
      -5.08307074]
     [-1.44470429 2.10092663 -9.13364161 ... 1.68764732
                                                         1.24891389
      -1.16454177]
     [-4.17360757 6.99745359 -9.6403615 ... 1.24891389
       0.38818873]
      \begin{bmatrix} -2.65854791 & 0.11198802 & -5.08307074 & \dots & -1.16454177 & 0.38818873 \end{bmatrix} 
       1.55084391]]
[9]: nercome_pos_count = 0
     nercome_neg_count = 0
     for row in nercome_rel_errors:
        for error in row:
             if error >= 0:
                 nercome_pos_count += 1
             elif error < 0:</pre>
                 nercome_neg_count += 1
```

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print(f"NERCOME overestimated {nercome_pos_count} elements")
      print(f"NERCOME underestimated {nercome_neg_count} elements")
      index_max_nercome = np.unravel_index(np.abs(nercome_rel_errors).argmax(),_
       →nercome_rel_errors.shape)
      print(f"Maximum relative error is {nercome_rel_errors[index_max_nercome]}")
      print(f"Maximum relative error index is {index_max_nercome}")
      print(f"NERCOME: {cov_nercome[index_max_nercome]}, real:__
       →{cov_real[index_max_nercome]}")
     NERCOME overestimated 7228 elements
     NERCOME underestimated 7172 elements
     Maximum relative error is -53538.754166466664
     Maximum relative error index is (25, 54)
     NERCOME: -1968.9173651550948, real: -0.03677486749440995
[10]: | sample_rel_errors = (cov_sample-cov_real)/np.abs(cov_real)
      print(sample rel errors)
     [-8.20827925e-02 -6.55513909e-01 -6.89950127e+00 ... -1.64606597e+00]
       -3.91684092e+00 -2.71670198e+00]
      [-6.55513909e-01 -1.29816282e-03 -5.16207807e-01 ... 1.57713825e+00
        6.04181999e+00 1.64306749e-01]
      [-6.89950127e+00 -5.16207807e-01 6.41871519e-02 ... -6.51507258e+00
       -7.68371504e+00 -4.62043849e+00]
      [-1.64606597e+00 1.57713825e+00 -6.51507258e+00 ... -2.69802957e-02
        1.92771215e-01 -9.99464014e-01]
      [-3.91684092e+00 6.04181999e+00 -7.68371504e+00 ... 1.92771215e-01
       -4.49197029e-02 -3.20388431e-01]
      [-2.71670198e+00 1.64306749e-01 -4.62043849e+00 ... -9.99464014e-01
       -3.20388431e-01 -2.27317615e-01]]
[11]: sample_pos_count = 0
      sample_neg_count = 0
      for row in sample_rel_errors:
          for error in row:
              if error >= 0:
                  sample_pos_count += 1
              elif error < 0:</pre>
                  sample neg count += 1
      print(f"Sample overestimated {sample pos count} elements")
      print(f"Sample underestimated {sample_neg_count} elements")
```

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index_max_sample = np.unravel_index(np.abs(sample_rel_errors).argmax(),_u
       ⇒sample_rel_errors.shape)
      print(f"Maximum relative error is {sample_rel_errors[index_max_sample]}")
      print(f"Maximum relative error index is {index_max_sample}")
      print(f"Sample: {cov_sample[index_max_sample]}, real:__
       →{cov_real[index_max_sample]}")
     Sample overestimated 6352 elements
     Sample underestimated 8048 elements
     Maximum relative error is 47402.094043586454
     Maximum relative error index is (80, 94)
     Sample: 631168.4055908481, real: -13.315481811674502
[12]: MSE_NERCOME = np.trace((cov_nercome-cov_real)@(cov_nercome-cov_real).T)
      MSE_sample = np.trace((cov_sample-cov_real)@(cov_sample-cov_real).T)
      print(f"MSE NERCOME: {MSE_NERCOME}")
      print(f"MSE sample: {MSE_sample}")
     MSE NERCOME: 2.814881523245139e+16
     MSE sample: 2.6448169256504176e+16
 []:
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