```
nan cm-2 s-1 TeV-1 nan nan
     amplitude 1.179e-12
                                                       False
     reference 1.000e+00
                           nan
                                          TeV nan nan
                                                        True
                                       TeV-1 nan nan
       lambda_ 5.668e-01
                                                      False
                           nan
         alpha 2.000e+00
                                                        True
                           nan
                                              nan nan
        name
                 value
                         error
                                   {\tt unit}
                                             min max frozen
         index 2.181e+00
                                              nan nan False
     amplitude 1.169e-12
                           nan cm-2 s-1 TeV-1 nan nan False
     reference 1.000e+00
                                        TeV nan nan
                           nan
       lambda_ 5.047e-01
                           nan
                                       TeV-1 nan nan False
         alpha 2.000e+00
                                              nan nan
                                                        True
                           nan
                 value
                         error
                                   \mathtt{unit}
                                              min max frozen
         index 2.317e+00
                           nan
                                              nan nan False
     amplitude 1.005e-12
                           nan cm-2 s-1 TeV-1 nan nan
                                                       False
     reference 1.000e+00
                                          TeV nan nan
                           nan
                                                        True
       lambda_ 3.616e-01
                                       TeV-1 nan nan
                                                      False
                           nan
         alpha 2.000e+00
                           nan
                                              nan nan
                                                        True
                 value
        name
                         error
                                   {\tt unit}
                                            min max frozen
         index 2.160e+00
                                              nan nan False
     amplitude 1.280e-12
                           nan cm-2 s-1 TeV-1 nan nan False
     reference 1.000e+00
                           nan
                                        TeV nan nan
       lambda_ 5.329e-01
                                       TeV-1 nan nan False
                           nan
         alpha 2.000e+00
                           nan
                                              nan nan
                                                        True
                 value
                         error
                                   {\tt unit}
                                              min max frozen
         index 2.301e+00
                                              nan nan False
     amplitude 1.151e-12
                           nan cm-2 s-1 TeV-1 nan nan
                                                       False
     reference 1.000e+00
                                          TeV nan nan
                           nan
       lambda_ 5.140e-01
                                       TeV-1 nan nan
                           nan
                                                       False
         alpha 2.000e+00
                                              nan nan
                           nan
        name
                 value
                                              min max frozen
                         error
                                   {\tt unit}
         index 2.171e+00
                                              nan nan False
     amplitude 1.243e-12
                           nan cm-2 s-1 TeV-1 nan nan False
     reference 1.000e+00
                           nan
                                         TeV nan nan
       lambda_ 5.495e-01
                                       TeV-1 nan nan False
                           nan
                                              nan nan
         alpha 2.000e+00
                                                        True
                           nan
[19]: index = [[0 for i in range(cols)] for j in range(rows)]
      amplitude = [[0 for i in range(cols)] for j in range(rows)]
      reference = [[0 for i in range(cols)] for j in range(rows)]
      lambda_ = [[0 for i in range(5)] for j in range(rows)]
      alpha = [[0 for i in range(5)] for j in range(rows)]
      covar = [[0 for i in range(cols)] for j in range(rows)]
```

```
for i in range(8):
          a = np.array([_["index"] for _ in res[i][0]])
          b = np.array([_["amplitude"] for _ in res[i][0]])
          c = np.array([_["reference"] for _ in res[i][0]])
          index[i][0]=a
          amplitude[i][0]=b
          reference[i][0]=c
      for j in range(5):
          for i in range(8):
              a = np.array([_["index"] for _ in res[i][j+1]])
              b = np.array([_["amplitude"] for _ in res[i][j+1]])
              c = np.array([_["reference"] for _ in res[i][j+1]])
              d = np.array([_["lambda_"] for _ in res[i][j+1]])
              e = np.array([_["alpha"] for _ in res[i][j+1]])
              index[i][j+1]=a
              amplitude[i][j+1]=b
              reference[i][j+1]=c
              lambda_[i][j]=d
              alpha[i][j]=e
      mu = [[0 for i in range(cols)] for j in range(rows)]
      sigma = [[0 for i in range(cols)] for j in range(rows)]
      for i in range(8):
          x = np.array([index[i][0], amplitude[i][0], reference[i][0]])
          covar[i][0]=np.cov(x)
      for j in range(5):
          for i in range(8):
              x = np.array([index[i][j+1], amplitude[i][j+1], reference[i][j+1],
       →lambda_[i][j], alpha[i][j]])
              covar[i][j+1]=np.cov(x)
[20]: def evaluate_err(self, covar, energy, epsilon=1e-4):
              p_cov = covar
              eps = np.sqrt(np.diag(covar)) * epsilon
              df_dp = self._evaluate_gradient(energy, eps)
              f_cov = df_dp.T @ p_cov @ df_dp
              f_err = np.sqrt(np.diagonal(f_cov))
              q = self(energy)
              return u.Quantity([q.value, f_err], unit=q.unit)
[21]: def plot_error(
              self,
              covar,
```

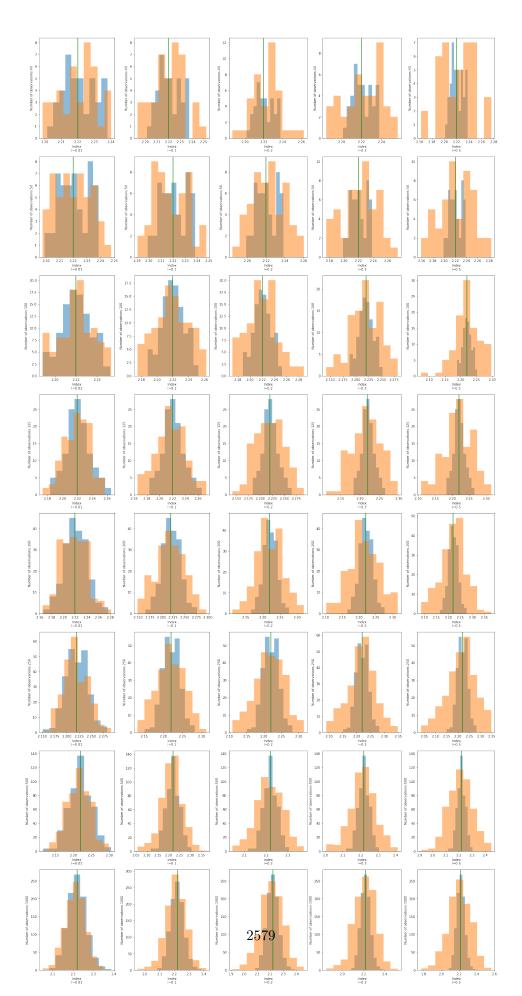
```
energy_range,
              ax=None,
              energy_unit="TeV",
              flux_unit="cm-2 s-1 TeV-1",
              energy_power=0,
              n_points=100,
              **kwargs,
          ):
              ax = plt.gca() if ax is None else ax
              kwargs.setdefault("facecolor", "black")
              kwargs.setdefault("alpha", 0.2)
              kwargs.setdefault("linewidth", 0)
              emin, emax = energy_range
              energy = MapAxis.from_energy_bounds(emin, emax, n_points, energy_unit).
       ⊶edges
              flux, flux_err = evaluate_err(self,covar,energy).to(flux_unit)
              y_lo = self._plot_scale_flux(energy, flux - flux_err, energy_power)
              y_hi = self._plot_scale_flux(energy, flux + flux_err, energy_power)
              where = (energy >= energy_range[0]) & (energy <= energy_range[1])</pre>
              ax.fill_between(energy.value, y_lo.value, y_hi.value, where=where, u
       →**kwargs)
              self._plot_format_ax(ax, energy, y_lo, energy_power)
              return ax
[22]: for j in range(6):
          print(f"model: {j}")
          for i in range(8):
              mu[i][j]=index[i][j].mean()
              sigma[i][j]=index[i][j].std()
              print(f"index: {index[i][j].mean()} += {index[i][j].std()}")
     model: 0
     index: 2.2208519415872168 += 0.00946474248505524
     index: 2.2204071142702646 += 0.010643317914159624
     index: 2.2203004860671913 += 0.014097739094078581
     index: 2.220852816179499 += 0.016172109015838997
     index: 2.2212194274890815 += 0.019215872375112058
     index: 2.2199563069339674 += 0.022096616282109238
     index: 2.2190755629793766 += 0.03098850638934652
     index: 2.219433716740164 += 0.045081260977561495
```

```
model: 1
index: 2.2204719698984876 += 0.011057911452127133
index: 2.2201525277145957 += 0.012376042811368266
index: 2.220766004292375 += 0.016254280452423428
index: 2.2192855660840825 += 0.016979350127733894
index: 2.2181942859260526 += 0.021566497637438602
index: 2.218340501343136 += 0.0246029745440613
index: 2.2141768111037137 += 0.03231075552018245
index: 2.2107577986756977 += 0.04844514153588179
model: 2
index: 2.2211847400271347 += 0.013650423195759882
index: 2.2186708810222173 += 0.013877953235843704
index: 2.2183847765756646 += 0.021281357410544443
index: 2.217352129681799 += 0.02251754427503892
index: 2.217878098387325 += 0.030017541358336335
index: 2.2180707820822607 += 0.03578842176931339
index: 2.2134213141794454 += 0.048266151351770606
index: 2.2096672550463956 += 0.06709402486086152
model: 3
index: 2.223099032157907 += 0.014582982639192716
index: 2.222378160459559 += 0.017317813920792062
index: 2.2193213565228627 += 0.02471830690583128
index: 2.2196622329825892 += 0.029492982027657766
index: 2.219270342305731 += 0.03810505221442646
index: 2.2164086393715294 += 0.042035907033355205
index: 2.214966998284206 += 0.061051815397505094
index: 2.210479984202571 += 0.08501738899244768
model: 4
index: 2.222540582907901 += 0.019103690191683814
index: 2.2226157841278593 += 0.022856886060554298
index: 2.2192448538730125 += 0.029345636261254127
index: 2.216770671204509 += 0.03677833344036589
index: 2.214961563969888 += 0.045036732115477464
index: 2.2177066375784715 += 0.04962759511556764
index: 2.2157036821177947 += 0.07098309361241216
index: 2.211728268618703 += 0.0990796020052331
model: 5
index: 2.2214952793651683 += 0.026250377207844712
index: 2.221494526385288 += 0.02940775753647641
index: 2.219397005708672 += 0.03882027909001693
index: 2.217771068100871 += 0.04167907227969314
index: 2.2204774474082605 += 0.05161966300111515
index: 2.2154705626624485 += 0.06378648415321286
index: 2.212090541476322 += 0.08989263357284584
index: 2.203583012383354 += 0.12668573135299435
```

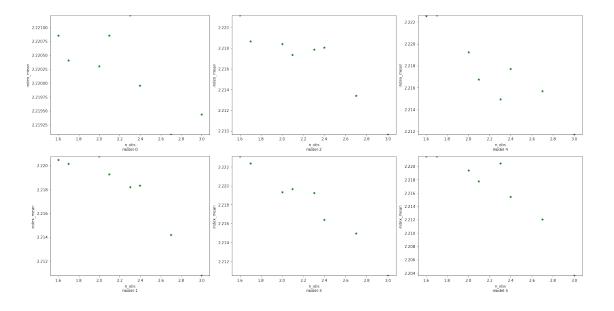
```
[23]: fig = plt.figure(figsize=[20,40],constrained_layout=True)
      import matplotlib.gridspec as gridspec
      gs0 = gridspec.GridSpec(1, 5, figure=fig)
      gs1 = gridspec.GridSpecFromSubplotSpec(8, 1, subplot_spec=gs0[0])
      for n in range(8):
          ax = fig.add_subplot(gs1[n])
          plt.hist(index[n][0], bins=10, alpha=0.5)
          plt.hist(index[n][1], bins=10, alpha=0.5)
          plt.axvline(x=model_simu.parameters["index"].value, color="green")
          plt.xlabel('Index\nl=0.01')
          plt.ylabel(f'Number of observations:{n_obs[n]}')
      gs2 = gridspec.GridSpecFromSubplotSpec(8, 1, subplot_spec=gs0[1])
      for n in range(8):
          ax = fig.add_subplot(gs2[n])
          plt.hist(index[n][0], bins=10, alpha=0.5)
          plt.hist(index[n][2], bins=10, alpha=0.5)
          plt.axvline(x=model_simu.parameters["index"].value, color="green")
          plt.xlabel('Index\nl=0.1')
          plt.ylabel(f'Number of observations:{n_obs[n]}')
      gs3 = gridspec.GridSpecFromSubplotSpec(8, 1, subplot_spec=gs0[2])
      for n in range(8):
          ax = fig.add_subplot(gs3[n])
          plt.hist(index[n][0], bins=10, alpha=0.5)
          plt.hist(index[n][3], bins=10, alpha=0.5)
          plt.axvline(x=model_simu.parameters["index"].value, color="green")
          plt.xlabel('Index\nl=0.2')
          plt.ylabel(f'Number of observations:{n_obs[n]}')
      gs4 = gridspec.GridSpecFromSubplotSpec(8, 1, subplot_spec=gs0[3])
      for n in range(8):
          ax = fig.add_subplot(gs4[n])
          plt.hist(index[n][0], bins=10, alpha=0.5)
          plt.hist(index[n][4], bins=10, alpha=0.5)
          plt.axvline(x=model_simu.parameters["index"].value, color="green")
          plt.xlabel('Index\nl=0.3')
          plt.ylabel(f'Number of observations:{n_obs[n]}')
      gs5 = gridspec.GridSpecFromSubplotSpec(8, 1, subplot_spec=gs0[4])
      for n in range(8):
          ax = fig.add_subplot(gs5[n])
          plt.hist(index[n][0], bins=10, alpha=0.5)
          plt.hist(index[n][5], bins=10, alpha=0.5)
```

```
plt.axvline(x=model_simu.parameters["index"].value, color="green")
plt.xlabel('Index\nl=0.5')
plt.ylabel(f'Number of observations:{n_obs[n]}')

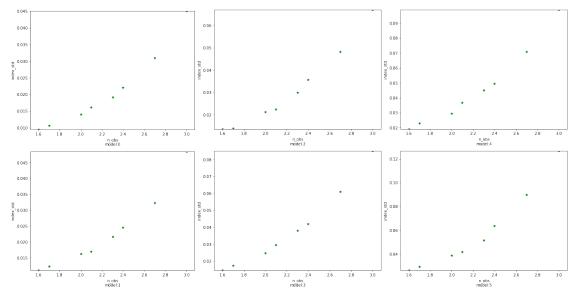
plt.show()
```



```
[24]: fig = plt.figure(figsize=[20,10],constrained_layout=True)
     import matplotlib.gridspec as gridspec
     gs0 = gridspec.GridSpec(1, 3, figure=fig)
     gs1 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[0])
     for n in range(2):
         ax = fig.add subplot(gs1[n])
         plt.scatter(np.log10(n_obs), [row[n] for row in mu], label= "stars", color=__
      ⇒"green",
                     marker= "*", s=30)
         plt.xlabel(f'n_obs\nmodel:{n}')
         plt.ylabel('index_mean')
         plt.ylim(min([row[n] for row in mu]),max([row[n] for row in mu]))
     gs2 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[1])
     for n in range(2):
         ax = fig.add_subplot(gs2[n])
         plt.scatter(np.log10(n_obs), [row[n+2] for row in mu], label= "stars", u
      marker= "*", s=30)
         plt.xlabel(f'n_obs\nmodel:{n+2}')
         plt.ylabel('index_mean')
         plt.ylim(min([row[n+2] for row in mu]),max([row[n+2] for row in mu]))
     gs3 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[2])
     for n in range(2):
         ax = fig.add_subplot(gs3[n])
         plt.scatter(np.log10(n_obs), [row[n+4] for row in mu], label= "stars", __
      marker= "*", s=30)
         plt.xlabel(f'n_obs\nmodel:{n+4}')
         plt.ylabel('index_mean')
         plt.ylim(min([row[n+4] for row in mu]),max([row[n+4] for row in mu]))
     plt.show()
```



```
[25]: fig = plt.figure(figsize=[20,10],constrained_layout=True)
     import matplotlib.gridspec as gridspec
     gs0 = gridspec.GridSpec(1, 3, figure=fig)
     gs1 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[0])
     for n in range(2):
         ax = fig.add_subplot(gs1[n])
         plt.scatter(np.log10(n_obs), [row[n] for row in sigma], label= "stars", __
      marker= "*", s=30)
         plt.xlabel(f'n_obs\nmodel:{n}')
         plt.ylabel('index_std')
         plt.ylim(min([row[n] for row in sigma]),max([row[n] for row in sigma]))
     gs2 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[1])
     for n in range(2):
         ax = fig.add_subplot(gs2[n])
         plt.scatter(np.log10(n_obs), [row[n+2] for row in sigma], label= "stars", __
      marker= "*", s=30)
         plt.xlabel(f'n_obs\nmodel:{n+2}')
         plt.ylabel('index_std')
         plt.ylim(min([row[n+2] for row in sigma]),max([row[n+2] for row in sigma]))
     gs3 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[2])
     for n in range(2):
```



```
[26]: for j in range(6):
    print(f"model: {j}")
    for i in range(8):
        mu[i][j]=amplitude[i][j].mean()
        sigma[i][j]=amplitude[i][j].std()
        print(f"amplitude: {amplitude[i][j].mean()} += {amplitude[i][j].std()}")

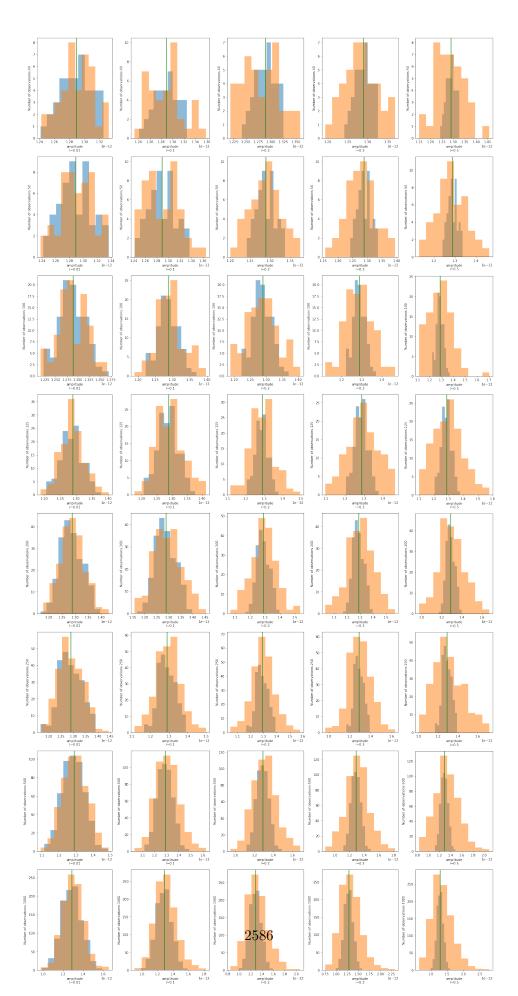
model: 0
```

amplitude: 1.2882109734645357e-12 += 2.020774416400257e-14 amplitude: 1.2887930617364896e-12 += 2.321243734503526e-14 amplitude: 1.2875057637201671e-12 += 2.943738660653206e-14 amplitude: 1.2855551328222816e-12 += 3.4620322478409695e-14 amplitude: 1.2852616877373924e-12 += 4.1590131875188173e-14 amplitude: 1.2870704254498414e-12 += 4.604036230233484e-14 amplitude: 1.2848003538014906e-12 += 6.382360983727948e-14 amplitude: 1.2834022687629476e-12 += 9.812239252655629e-14 model: 1 amplitude: 1.2887740733863474e-12 += 2.1624008113134554e-14 amplitude: 1.290197249320048e-12 += 2.3665712755630992e-14

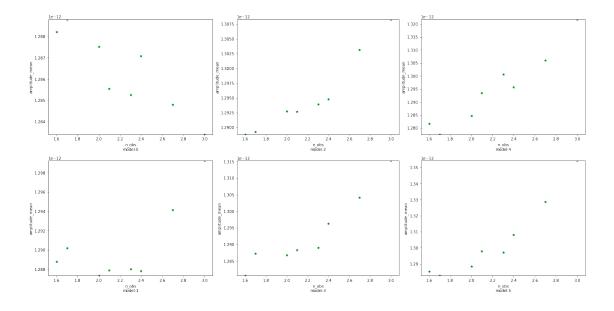
```
amplitude: 1.2873240819719735e-12 += 3.2647047697068356e-14
     amplitude: 1.287907206311591e-12 += 3.64426974840847e-14
     amplitude: 1.288016568088648e-12 += 4.6423404433813026e-14
     amplitude: 1.2878012385332741e-12 += 4.866857799313919e-14
     amplitude: 1.2941386042810284e-12 += 6.980329125688868e-14
     amplitude: 1.299253568211679e-12 += 1.032452149663184e-13
     model: 2
     amplitude: 1.288833703100092e-12 += 2.961453439686997e-14
     amplitude: 1.2893006402619744e-12 += 2.8680818735483707e-14
     amplitude: 1.2927832874446586e-12 += 3.988864742203177e-14
     amplitude: 1.2927113396963073e-12 += 4.504148198217889e-14
     amplitude: 1.2939662295353098e-12 += 5.825767301477067e-14
     amplitude: 1.2947862319018303e-12 += 6.618781500750589e-14
     amplitude: 1.3031686219183516e-12 += 9.697874818989171e-14
     amplitude: 1.3082828025427663e-12 += 1.3892983026905873e-13
     model: 3
     amplitude: 1.2805526881050745e-12 += 3.574018932168858e-14
     amplitude: 1.2872566281400055e-12 += 3.844094690616137e-14
     amplitude: 1.286795800901381e-12 += 4.978203409314228e-14
     amplitude: 1.2882630704524287e-12 += 6.442621190119476e-14
     amplitude: 1.2890386543647814e-12 += 8.12412098445936e-14
     amplitude: 1.296306142469232e-12 += 8.772249466134022e-14
     amplitude: 1.3041335523312662e-12 += 1.240115995567233e-13
     amplitude: 1.3152171172761559e-12 += 1.8316141365511399e-13
     model: 4
     amplitude: 1.281598735194535e-12 += 4.1968762789464806e-14
     amplitude: 1.2775118349799447e-12 += 4.9074789194400446e-14
     amplitude: 1.2846474844088973e-12 += 6.617702199344377e-14
     amplitude: 1.2934970969874811e-12 += 8.226292423276059e-14
     amplitude: 1.3006942825709336e-12 += 1.0004240556633877e-13
     amplitude: 1.2956693454424653e-12 += 1.0648717184660167e-13
     amplitude: 1.3060211186389223e-12 += 1.5765578083564723e-13
     amplitude: 1.3217526457383794e-12 += 2.287718530959276e-13
     model: 5
     amplitude: 1.2852430965817646e-12 += 6.24353802813177e-14
     amplitude: 1.2825788705042499e-12 += 7.073938775851406e-14
     amplitude: 1.2884404569577469e-12 += 9.577571033132101e-14
     amplitude: 1.2980170584135136e-12 += 9.830052853675669e-14
     amplitude: 1.2971893243690907e-12 += 1.2750395315534658e-13
     amplitude: 1.308078615540899e-12 += 1.4960139056854337e-13
     amplitude: 1.3287393295332176e-12 += 2.1348786203556218e-13
     amplitude: 1.3542334224473133e-12 += 3.2352153412172364e-13
[27]: fig = plt.figure(figsize=[20,40],constrained_layout=True)
      import matplotlib.gridspec as gridspec
```

```
gs0 = gridspec.GridSpec(1, 5, figure=fig)
gs1 = gridspec.GridSpecFromSubplotSpec(8, 1, subplot_spec=gs0[0])
for n in range(8):
   ax = fig.add_subplot(gs1[n])
   plt.hist(amplitude[n][0], bins=10, alpha=0.5)
   plt.hist(amplitude[n][1], bins=10, alpha=0.5)
   plt.axvline(x=model_simu.parameters["amplitude"].value, color="green")
   plt.xlabel('amplitude\nl=0.01')
   plt.ylabel(f'Number of observations:{n_obs[n]}')
gs2 = gridspec.GridSpecFromSubplotSpec(8, 1, subplot_spec=gs0[1])
for n in range(8):
   ax = fig.add_subplot(gs2[n])
   plt.hist(amplitude[n][0], bins=10, alpha=0.5)
   plt.hist(amplitude[n][2], bins=10, alpha=0.5)
   plt.axvline(x=model_simu.parameters["amplitude"].value, color="green")
   plt.xlabel('amplitude\nl=0.1')
   plt.ylabel(f'Number of observations:{n_obs[n]}')
gs3 = gridspec.GridSpecFromSubplotSpec(8, 1, subplot_spec=gs0[2])
for n in range(8):
   ax = fig.add_subplot(gs3[n])
   plt.hist(amplitude[n][0], bins=10, alpha=0.5)
   plt.hist(amplitude[n][3], bins=10, alpha=0.5)
   plt.axvline(x=model_simu.parameters["amplitude"].value, color="green")
   plt.xlabel('amplitude\nl=0.2')
   plt.ylabel(f'Number of observations:{n obs[n]}')
gs4 = gridspec.GridSpecFromSubplotSpec(8, 1, subplot_spec=gs0[3])
for n in range(8):
   ax = fig.add_subplot(gs4[n])
   plt.hist(amplitude[n][0], bins=10, alpha=0.5)
   plt.hist(amplitude[n][4], bins=10, alpha=0.5)
   plt.axvline(x=model_simu.parameters["amplitude"].value, color="green")
   plt.xlabel('amplitude\nl=0.3')
   plt.ylabel(f'Number of observations:{n_obs[n]}')
gs5 = gridspec.GridSpecFromSubplotSpec(8, 1, subplot_spec=gs0[4])
for n in range(8):
   ax = fig.add subplot(gs5[n])
   plt.hist(amplitude[n][0], bins=10, alpha=0.5)
   plt.hist(amplitude[n][5], bins=10, alpha=0.5)
   plt.axvline(x=model_simu.parameters["amplitude"].value, color="green")
   plt.xlabel('amplitude\nl=0.5')
   plt.ylabel(f'Number of observations:{n_obs[n]}')
```

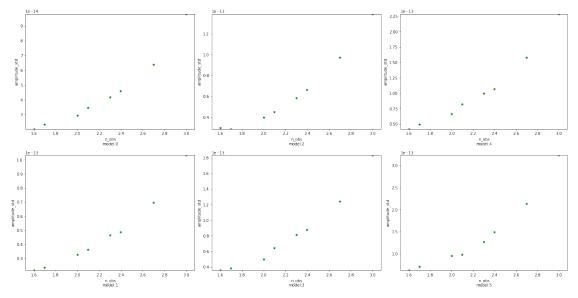
plt.show()



```
[28]: fig = plt.figure(figsize=[20,10],constrained_layout=True)
     import matplotlib.gridspec as gridspec
     gs0 = gridspec.GridSpec(1, 3, figure=fig)
     gs1 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[0])
     for n in range(2):
         ax = fig.add subplot(gs1[n])
         plt.scatter(np.log10(n_obs), [row[n] for row in mu], label= "stars", color=__
      ⇒"green",
                     marker= "*", s=30)
         plt.xlabel(f'n_obs\nmodel:{n}')
         plt.ylabel('amplitude_mean')
         plt.ylim(min([row[n] for row in mu]),max([row[n] for row in mu]))
     gs2 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[1])
     for n in range(2):
         ax = fig.add_subplot(gs2[n])
         plt.scatter(np.log10(n_obs), [row[n+2] for row in mu], label= "stars", u
      marker= "*", s=30)
         plt.xlabel(f'n_obs\nmodel:{n+2}')
         plt.ylabel('amplitude_mean')
         plt.ylim(min([row[n+2] for row in mu]),max([row[n+2] for row in mu]))
     gs3 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[2])
     for n in range(2):
         ax = fig.add_subplot(gs3[n])
         plt.scatter(np.log10(n_obs), [row[n+4] for row in mu], label= "stars", __
      marker= "*", s=30)
         plt.xlabel(f'n_obs\nmodel:{n+4}')
         plt.ylabel('amplitude_mean')
         plt.ylim(min([row[n+4] for row in mu]),max([row[n+4] for row in mu]))
     plt.show()
```



```
[29]: fig = plt.figure(figsize=[20,10],constrained_layout=True)
     import matplotlib.gridspec as gridspec
     gs0 = gridspec.GridSpec(1, 3, figure=fig)
     gs1 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[0])
     for n in range(2):
         ax = fig.add_subplot(gs1[n])
         plt.scatter(np.log10(n_obs), [row[n] for row in sigma], label= "stars", __
      marker= "*", s=30)
         plt.xlabel(f'n_obs\nmodel:{n}')
         plt.ylabel('amplitude_std')
         plt.ylim(min([row[n] for row in sigma]),max([row[n] for row in sigma]))
     gs2 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[1])
     for n in range(2):
         ax = fig.add_subplot(gs2[n])
         plt.scatter(np.log10(n_obs), [row[n+2] for row in sigma], label= "stars", __
      marker= "*", s=30)
         plt.xlabel(f'n_obs\nmodel:{n+2}')
         plt.ylabel('amplitude_std')
         plt.ylim(min([row[n+2] for row in sigma]),max([row[n+2] for row in sigma]))
     gs3 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[2])
     for n in range(2):
```



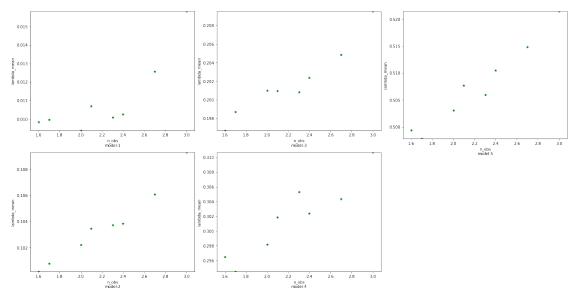
```
[30]: for j in range(5):
    print(f"model: {j}")
    for i in range(8):
        mu[i][j]=lambda_[i][j].mean()
        sigma[i][j]=lambda_[i][j].std()
        print(f"lambda_: {lambda_[i][j].mean()} += {lambda_[i][j].std()}")
```

model: 0
lambda_: 0.00983328872701745 += 0.003527412815651705
lambda_: 0.00997096588039571 += 0.003855226271571548
lambda_: 0.009380252964157535 += 0.00519607994559184
lambda_: 0.01069061050567699 += 0.005867950169742963
lambda_: 0.01008256122878477 += 0.007212265102101397
lambda_: 0.010256444061833998 += 0.007588274308204233
lambda_: 0.012567652159184253 += 0.010904987005412904
lambda_: 0.015828571114078694 += 0.016919554721768242
model: 1
lambda_: 0.10015432945628366 += 0.0066480054181108465
lambda_: 0.10078827921563097 += 0.006449629883359707

```
lambda_: 0.10344886668194597 += 0.013432590106282118
     lambda_: 0.10373798639547463 += 0.01640438049913423
     lambda_: 0.10383728053928191 += 0.01839816538948341
     lambda : 0.10607361404282202 += 0.027229784201596493
     lambda_: 0.1092707065622827 += 0.03675385733319002
     model: 2
     lambda : 0.1967069864590812 += 0.010744366602316709
     lambda : 0.19869165590213667 += 0.013229014698656682
     lambda_: 0.20099872976971228 += 0.017449749215610166
     lambda_: 0.2009619078702176 += 0.02269908573695963
     lambda_: 0.20083837141583402 += 0.025916384082007707
     lambda_: 0.2023882407275771 += 0.030565647809145686
     lambda_: 0.2048623674512616 += 0.04117116322471812
     lambda_: 0.20954778762943943 += 0.06034534310983556
     model: 3
     lambda_: 0.2964729205851758 += 0.014721944589363148
     lambda_: 0.29447525989345485 += 0.01685203921803965
     lambda : 0.29817104986664195 += 0.027015407948679744
     lambda_: 0.30185892306290885 += 0.030489761434818297
     lambda : 0.3052778294324254 += 0.03695667157246595
     lambda : 0.3024098305014937 += 0.03962871920754218
     lambda_: 0.30436087505589615 += 0.05804036947085071
     lambda_: 0.31065997881282587 += 0.08196051750382335
     model: 4
     lambda_: 0.4993696065408228 += 0.026190993490909777
     lambda_: 0.49778683814020547 += 0.029522341153212735
     lambda_: 0.5030743790002709 += 0.038467654459497
     lambda : 0.5076719641434785 += 0.04428667967358409
     lambda_: 0.5059721243539873 += 0.05600715517347292
     lambda_: 0.5105196288691412 += 0.070808907235574
     lambda_: 0.5148255650991596 += 0.0966096298646213
     lambda_: 0.5214624796108971 += 0.13687383699270922
[31]: fig = plt.figure(figsize=[20,10],constrained_layout=True)
      import matplotlib.gridspec as gridspec
      gs0 = gridspec.GridSpec(1, 3, figure=fig)
      gs1 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[0])
      for n in range(2):
          ax = fig.add_subplot(gs1[n])
          plt.scatter(np.log10(n_obs), [row[n] for row in mu], label= "stars", color=__
       marker= "*", s=30)
          plt.xlabel(f'n obs\nmodel:{n+1}')
```

lambda_: 0.1022222867459217 += 0.011100524725229686

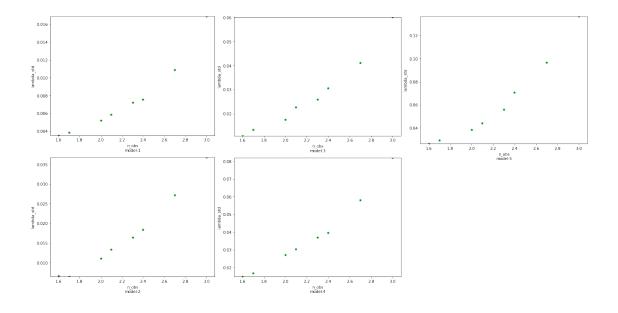
```
plt.ylabel('lambda_mean')
   plt.ylim(min([row[n] for row in mu]),max([row[n] for row in mu]))
gs2 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[1])
for n in range(2):
   ax = fig.add_subplot(gs2[n])
   plt.scatter(np.log10(n_obs), [row[n+2] for row in mu], label= "stars", __
marker= "*", s=30)
   plt.xlabel(f'n_obs\nmodel:{n+3}')
   plt.ylabel('lambda_mean')
   plt.ylim(min([row[n+2] for row in mu]),max([row[n+2] for row in mu]))
gs3 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[2])
for n in range(1):
   ax = fig.add_subplot(gs3[n])
   plt.scatter(np.log10(n_obs), [row[n+4] for row in mu], label= "stars", __
marker= "*", s=30)
   plt.xlabel(f'n_obs\nmodel:{n+5}')
   plt.ylabel('lambda_mean')
   plt.ylim(min([row[n+4] for row in mu]),max([row[n+4] for row in mu]))
plt.show()
```



```
[32]: fig = plt.figure(figsize=[20,10],constrained_layout=True)

import matplotlib.gridspec as gridspec
```

```
gs0 = gridspec.GridSpec(1, 3, figure=fig)
gs1 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[0])
for n in range(2):
   ax = fig.add_subplot(gs1[n])
   plt.scatter(np.log10(n_obs), [row[n] for row in sigma], label= "stars", __
marker= "*", s=30)
   plt.xlabel(f'n_obs\nmodel:{n+1}')
   plt.ylabel('lambda_std')
   plt.ylim(min([row[n] for row in sigma]),max([row[n] for row in sigma]))
gs2 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[1])
for n in range(2):
   ax = fig.add_subplot(gs2[n])
   plt.scatter(np.log10(n_obs), [row[n+2] for row in sigma], label= "stars", u
marker= "*", s=30)
   plt.xlabel(f'n_obs\nmodel:{n+3}')
   plt.ylabel('lambda_std')
   plt.ylim(min([row[n+2] for row in sigma]),max([row[n+2] for row in sigma]))
gs3 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[2])
for n in range(1):
   ax = fig.add_subplot(gs3[n])
   plt.scatter(np.log10(n_obs), [row[n+4] for row in sigma], label= "stars", __
marker= "*", s=30)
   plt.xlabel(f'n_obs\nmodel:{n+5}')
   plt.ylabel('lambda_std')
   plt.ylim(min([row[n+4] for row in sigma]),max([row[n+4] for row in sigma]))
plt.show()
```



```
[33]: sim = [[0 for i in range(cols)] for j in range(rows)]
      for i in range(8):
          s = PowerLawSpectralModel(
              index=index[i][0].mean(),
              amplitude=amplitude[i][0].mean() * u.Unit("cm-2 s-1 TeV-1"),
              reference=1 * u.TeV,
          print(s)
          sim[i][0]=s
      for j in range(5):
          for i in range(8):
              s = ExpCutoffPowerLawSpectralModel(
                  index=index[i][j+1].mean(),
                  amplitude=amplitude[i][j+1].mean() * u.Unit("cm-2 s-1 TeV-1"),
                  reference=1 * u.TeV,
                  lambda_=lambda_[i][j].mean() * u.Unit("TeV-1"),
                  alpha = 2,
              print(s)
              sim[i][j+1]=s
```

PowerLawSpectralModel

```
        name
        value
        error
        unit
        min
        max
        frozen

        index
        2.221e+00
        nan
        nan
        nan
        nan
        False

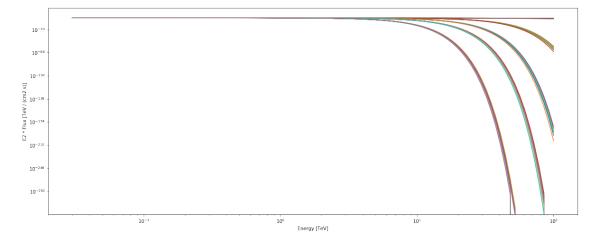
        amplitude
        1.288e-12
        nan
        cm-2
        s-1
        TeV-1
        nan
        nan
        True
```

```
value
                   error
                             unit
                                      min max frozen
  name
   index 2.215e+00
                                       nan nan False
                    nan
amplitude 1.308e-12
                    nan cm-2 s-1 TeV-1 nan nan False
reference 1.000e+00
                    nan
                                   TeV nan nan
                                                True
  lambda_ 5.105e-01
                    nan
                               TeV-1 nan nan False
   alpha 2.000e+00
                    nan
                                      nan nan
ExpCutoffPowerLawSpectralModel
```

name	value	error	unit		min	max	frozen	
index	2.212e+00	nan				nan	nan	False
amplitude	1.329e-12	nan	cm-2	s-1	TeV-1	nan	nan	False
reference	1.000e+00	nan			TeV	nan	nan	True
lambda_	5.148e-01	nan			TeV-1	nan	nan	False
alpha	2.000e+00	nan				nan	nan	True
ExpCutoffPowerLawSpectralModel								

value unit min max frozen nameerror index 2.204e+00 nan nan False amplitude 1.354e-12 nan cm-2 s-1 TeV-1 nan nan False reference 1.000e+00 nan TeV nan nan True lambda_ 5.215e-01 TeV-1 nan nan False nan alpha 2.000e+00 nan nan nan True

```
plt.figure(figsize=[20,8])
energy_range = [0.03, 100] * u.TeV
for j in range(6):
    for i in range(8):
        sim[i][j].plot(energy_range=energy_range, energy_power=2)
plt.show
plt.savefig('energy_lightcurve')
```

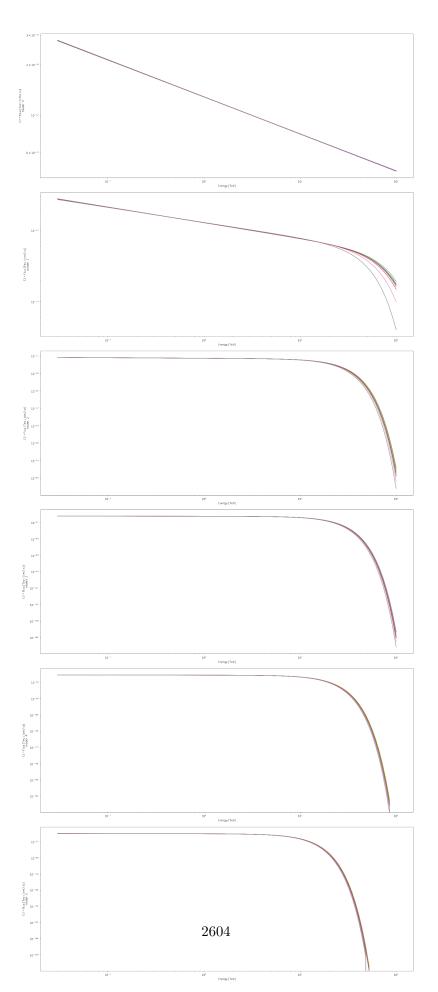


```
[35]: fig = plt.figure(figsize=[20,48],constrained_layout=True)
    import matplotlib.gridspec as gridspec

gs0 = gridspec.GridSpec(1, 1, figure=fig)

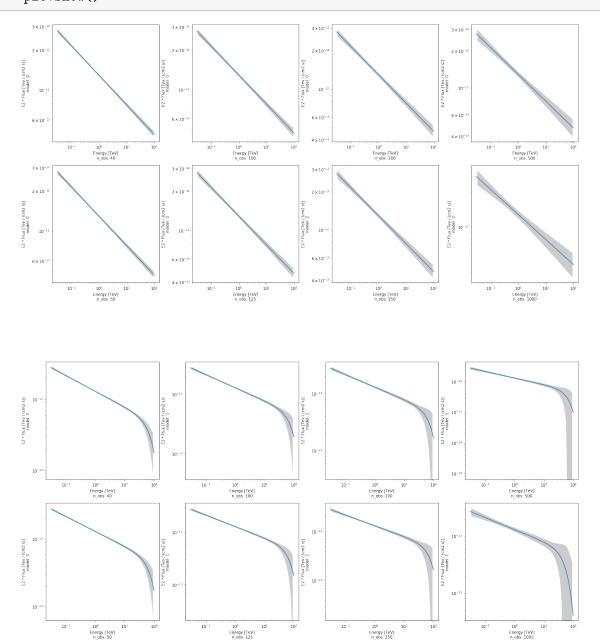
gs1 = gridspec.GridSpecFromSubplotSpec(6, 1, subplot_spec=gs0[0])
    for n in range(6):
        ax = fig.add_subplot(gs1[n])
        for i in range(8):
            sim[i][n].plot(energy_range=energy_range, energy_power=2)
        plt.ylabel(f'E2 * Flux [Tev / (cm2 s)]\nmodel: {n}')

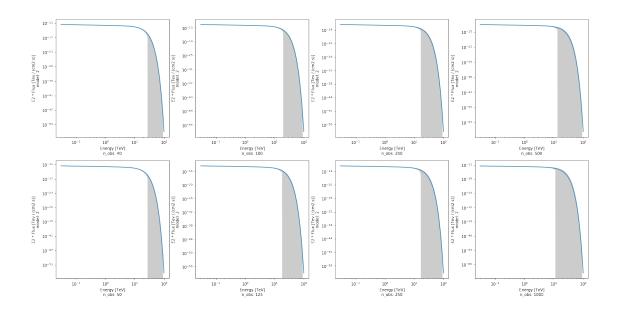
plt.show()
    plt.savefig('energy_lightcurve')
```

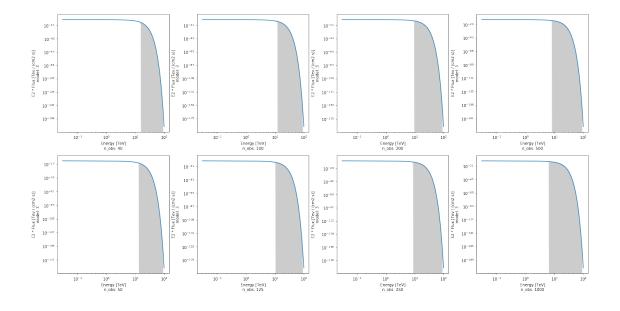


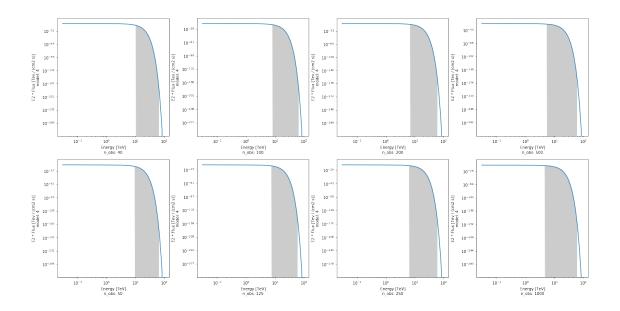
```
[36]: for i in range(6):
          fig = plt.figure(figsize=[20,10],constrained_layout=True)
          import matplotlib.gridspec as gridspec
          gs0 = gridspec.GridSpec(1, 4, figure=fig)
          gs1 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[0])
          for n in range(2):
              ax = fig.add_subplot(gs1[n])
              sim[n][i].plot(energy_range=energy_range, energy_power=2)
              plot_error(self=sim[n][i], covar=covar[n][i],energy_range=energy_range,_u
       →energy_power=2)
              plt.xlabel(f'Energy [TeV]\nn_obs: {n_obs[n]}')
              plt.ylabel(f'E2 * Flux [Tev / (cm2 s)]\nmodel: {i}')
          gs2 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[1])
          for n in range(2):
              ax = fig.add_subplot(gs2[n])
              sim[n+2][i].plot(energy_range=energy_range, energy_power=2)
              plot_error(self=sim[n+2][i],__
       →covar=covar[n+2][i],energy_range=energy_range, energy_power=2)
              plt.xlabel(f'Energy [TeV]\nn_obs: {n_obs[n+2]}')
              plt.ylabel(f'E2 * Flux [Tev / (cm2 s)]\nmodel: {i}')
          gs3 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[2])
          for n in range(2):
              ax = fig.add_subplot(gs3[n])
              sim[n+4][i].plot(energy_range=energy_range, energy_power=2)
              plot_error(self=sim[n+4][i],_u
       \hookrightarrowcovar=covar[n+4][i],energy_range=energy_range, energy_power=2)
              plt.xlabel(f'Energy [TeV]\nn_obs: {n_obs[n+4]}')
              plt.ylabel(f'E2 * Flux [Tev / (cm2 s)]\nmodel: {i}')
          gs4 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[3])
          for n in range(2):
              ax = fig.add_subplot(gs4[n])
              sim[n+6][i].plot(energy_range=energy_range, energy_power=2)
              plot_error(self=sim[n+6][i],__
       →covar=covar[n+6][i],energy_range=energy_range, energy_power=2)
              plt.xlabel(f'Energy [TeV]\nn_obs: {n_obs[n+6]}')
              plt.ylabel(f'E2 * Flux [Tev / (cm2 s)]\nmodel: {i}')
```

plt.show()



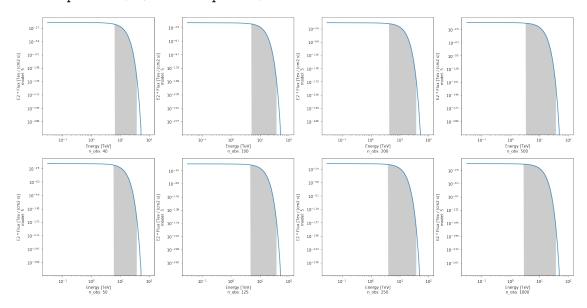






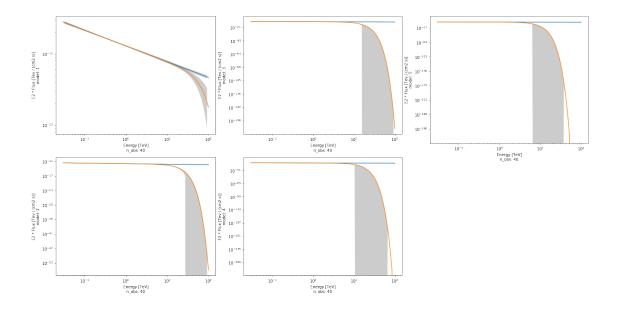
/home/rishank/anaconda2/envs/gammapy-0.15/lib/python3.6/site-packages/matplotlib/ticker.py:1123: RuntimeWarning: divide by zero encountered in double_scalars

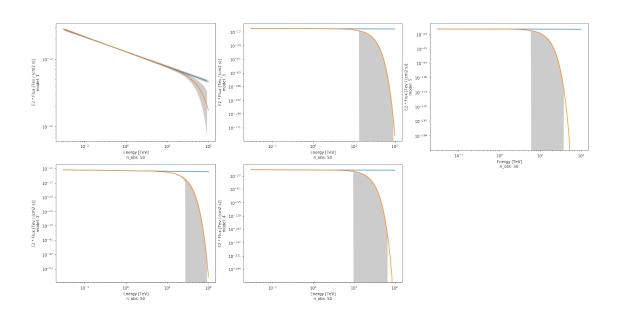
coeff = np.round(x / b ** exponent)

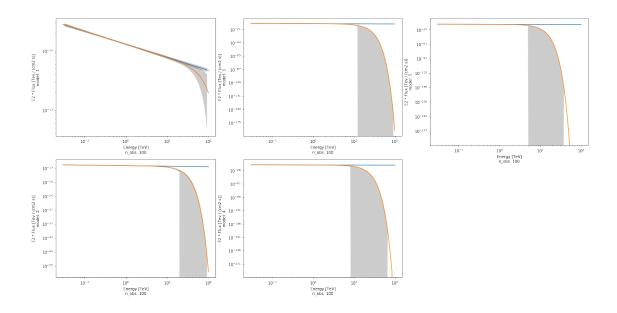


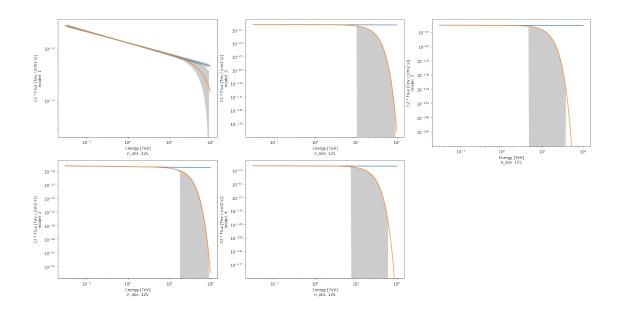
```
[37]: for i in range(8):
    fig = plt.figure(figsize=[20,10],constrained_layout=True)
    import matplotlib.gridspec as gridspec
```

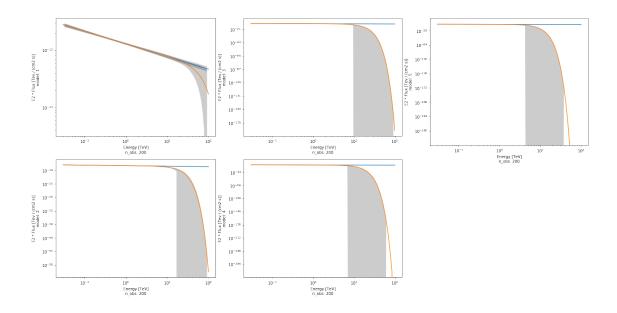
```
gs0 = gridspec.GridSpec(1, 3, figure=fig)
  gs1 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[0])
  for n in range(2):
      ax = fig.add_subplot(gs1[n])
       sim[i][0].plot(energy_range=energy_range, energy_power=2)
       plot_error(self=sim[i][0], covar=covar[i][0],energy_range=energy_range,u
→energy_power=2)
       sim[i][n+1].plot(energy_range=energy_range, energy_power=2)
       plot_error(self=sim[i][n+1],__
→covar=covar[i][n+1],energy_range=energy_range, energy_power=2)
       plt.xlabel(f'Energy [TeV]\nn obs: {n obs[i]}')
      plt.ylabel(f'E2 * Flux [Tev / (cm2 s)]\nmodel: {n+1}')
  gs2 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[1])
  for n in range(2):
       ax = fig.add_subplot(gs2[n])
       sim[i][0].plot(energy_range=energy_range, energy_power=2)
      plot_error(self=sim[i][0], covar=covar[i][0],energy_range=energy_range,u
→energy_power=2)
       sim[i][n+3].plot(energy_range=energy_range, energy_power=2)
       plot_error(self=sim[i][n+3],__
→covar=covar[i][n+3],energy_range=energy_range, energy_power=2)
       plt.xlabel(f'Energy [TeV]\nn_obs: {n_obs[i]}')
      plt.ylabel(f'E2 * Flux [Tev / (cm2 s)]\nmodel: {n+3}')
  gs3 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[2])
  for n in range(1):
       ax = fig.add_subplot(gs3[n])
       sim[i][0].plot(energy_range=energy_range, energy_power=2)
       plot_error(self=sim[i][0], covar=covar[i][0],energy_range=energy_range,u
→energy_power=2)
      sim[i][n+5].plot(energy_range=energy_range, energy_power=2)
      plot_error(self=sim[i][n+5],_u
→covar=covar[i][n+5],energy_range=energy_range, energy_power=2)
      plt.xlabel(f'Energy [TeV]\nn_obs: {n_obs[i]}')
      plt.ylabel(f'E2 * Flux [Tev / (cm2 s)]\nmodel: {n+5}')
  plt.show()
```

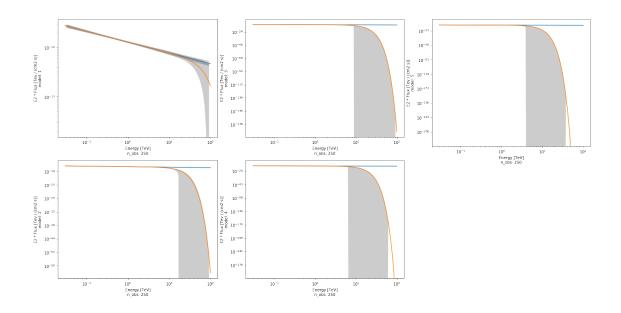


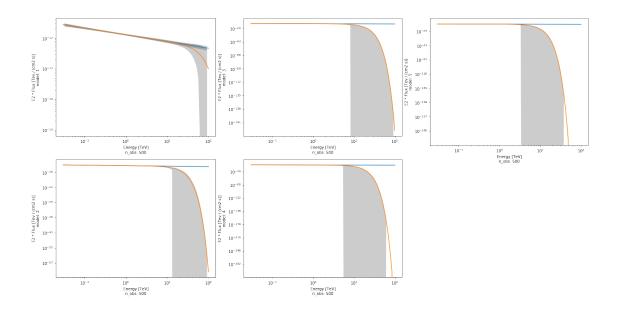


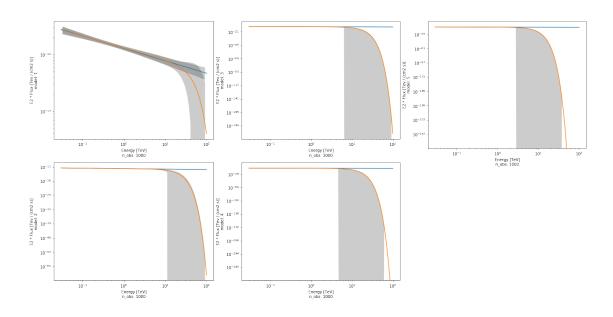












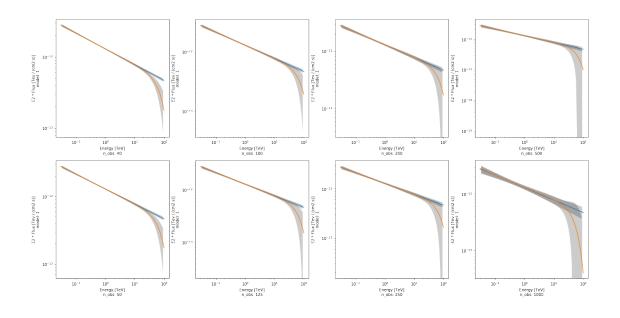
```
for i in range(5):
    fig = plt.figure(figsize=[20,10],constrained_layout=True)

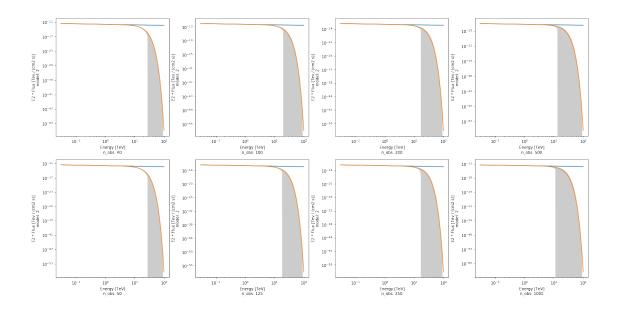
import matplotlib.gridspec as gridspec

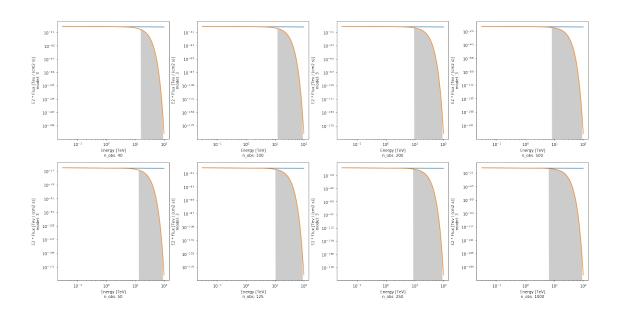
gs0 = gridspec.GridSpec(1, 4, figure=fig)

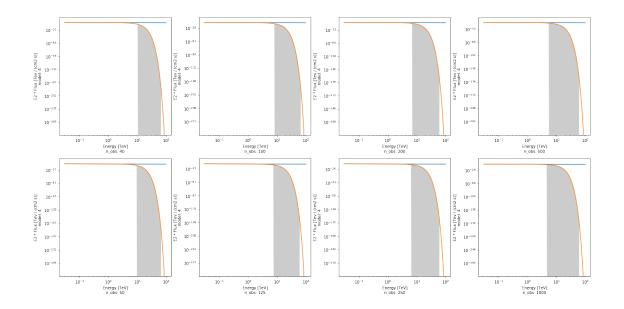
gs1 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[0])
    for n in range(2):
        ax = fig.add_subplot(gs1[n])
```

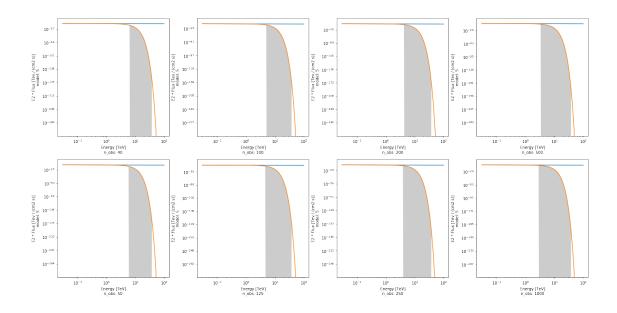
```
sim[n][0].plot(energy_range=energy_range, energy_power=2)
      plot_error(self=sim[n][0], covar=covar[n][0],energy_range=energy_range,u
→energy_power=2)
       sim[n][i+1].plot(energy_range=energy_range, energy_power=2)
       plot_error(self=sim[n][i+1],__
→covar=covar[n][i+1],energy_range=energy_range, energy_power=2)
       plt.xlabel(f'Energy [TeV]\nn_obs: {n_obs[n]}')
      plt.ylabel(f'E2 * Flux [Tev / (cm2 s)]\nmodel: {i+1}')
  gs2 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[1])
  for n in range(2):
       ax = fig.add_subplot(gs2[n])
       sim[n+2][0].plot(energy_range=energy_range, energy_power=2)
       plot_error(self=sim[n+2][0],_
→covar=covar[n+2][0],energy_range=energy_range, energy_power=2)
       sim[n+2][i+1].plot(energy_range=energy_range, energy_power=2)
      plot_error(self=sim[n+2][i+1],_u
→covar=covar[n+2][i+1],energy_range=energy_range, energy_power=2)
       plt.xlabel(f'Energy [TeV]\nn_obs: {n_obs[n+2]}')
      plt.ylabel(f'E2 * Flux [Tev / (cm2 s)]\nmodel: {i+1}')
  gs3 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[2])
  for n in range(2):
       ax = fig.add_subplot(gs3[n])
       sim[n+4][0].plot(energy_range=energy_range, energy_power=2)
       plot_error(self=sim[n+4][0],__
→covar=covar[n+4][0],energy_range=energy_range, energy_power=2)
       sim[n+4][i+1].plot(energy_range=energy_range, energy_power=2)
      plot_error(self=sim[n+4][i+1],__
→covar=covar[n+4][i+1],energy_range=energy_range, energy_power=2)
      plt.xlabel(f'Energy [TeV]\nn_obs: {n_obs[n+4]}')
      plt.ylabel(f'E2 * Flux [Tev / (cm2 s)]\nmodel: {i+1}')
  gs4 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[3])
  for n in range(2):
       ax = fig.add_subplot(gs4[n])
       sim[n+6][0].plot(energy_range=energy_range, energy_power=2)
      plot_error(self=sim[n+6][0],__
→covar=covar[n+6][0],energy_range=energy_range, energy_power=2)
       sim[n+6][i+1].plot(energy_range=energy_range, energy_power=2)
      plot_error(self=sim[n+6][i+1],__
→covar=covar[n+6][i+1],energy_range=energy_range, energy_power=2)
      plt.xlabel(f'Energy [TeV]\nn_obs: {n_obs[n+6]}')
      plt.ylabel(f'E2 * Flux [Tev / (cm2 s)]\nmodel: {i+1}')
  plt.show()
```











```
[39]: %%time
      res_1 = [[0 for i in range(cols)] for j in range(rows)]
      minuit_opts = {"tol": 0.001, "strategy": 1}
      for i in range(8):
          results_1 = []
          for dataset in datas[i][0]:
              dataset.models = model[i][0].copy()
              fit_1 = Fit([dataset])
              result_1 = fit_1.run(optimize_opts=minuit_opts)
              results_1.append(
                  {
                      "index": result_1.parameters["index"].value,
                      "amplitude": result_1.parameters["amplitude"].value,
                      "reference":result_1.parameters["reference"].value,
                      "covariance": result_1.parameters.covariance,
                  }
              print(result_1.parameters.to_table())
          res_1[i][0]=results_1
```

```
name
            value
                                    unit
                                               min max frozen
    index 2.210e+00 9.028e-03
                                               nan nan
                                                        False
amplitude 1.318e-12\ 1.964e-14\ cm-2\ s-1\ TeV-1\ nan\ nan
                                                        False
reference 1.000e+00 0.000e+00
                                           TeV nan nan
                                                         True
            value
   name
                       error
                                    unit
                                               min max frozen
    index 2.222e+00 9.318e-03
                                               nan nan False
amplitude 1.258e-12 1.930e-14 cm-2 s-1 TeV-1 nan nan False
```

value error unit min max frozen _____ ___ ____ index 2.220e+00 nan nan nan False amplitude 1.297e-12 nan cm-2 s-1 TeV-1 nan nan False reference 1.000e+00 nan TeV nan nan True TeV-1 nan nan False lambda 5.060e-01 nan alpha 2.000e+00 nan nan nan True ExpCutoffPowerLawSpectralModel namevalue error unit min max frozen ----- ---- ----index 2.215e+00 nan nan nan False amplitude 1.308e-12 nan cm-2 s-1 TeV-1 nan nan False reference 1.000e+00 nan TeV nan nan True nan TeV-1 nan nan False lambda_ 5.105e-01 alpha 2.000e+00 nan nan nan True ExpCutoffPowerLawSpectralModel value error unit min max frozen name ----index 2.212e+00 nan nan nan False amplitude 1.329e-12 nan cm-2 s-1 TeV-1 nan nan False reference 1.000e+00 TeV nan nan nan True lambda_ 5.148e-01 TeV-1 nan nan False nan alpha 2.000e+00 nan nan nan True ExpCutoffPowerLawSpectralModel name value error unit min max frozen _____ ___ ____ index 2.204e+00 nan nan nan False amplitude 1.354e-12 nan cm-2 s-1 TeV-1 nan nan False reference 1.000e+00 nan TeV nan nan True TeV-1 nan nan False lambda_ 5.215e-01 nan alpha 2.000e+00 nan nan True nan [43]: for i in range(6): fig = plt.figure(figsize=[20,10],constrained_layout=True) import matplotlib.gridspec as gridspec gs0 = gridspec.GridSpec(1, 4, figure=fig) gs1 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[0]) for n in range(2):

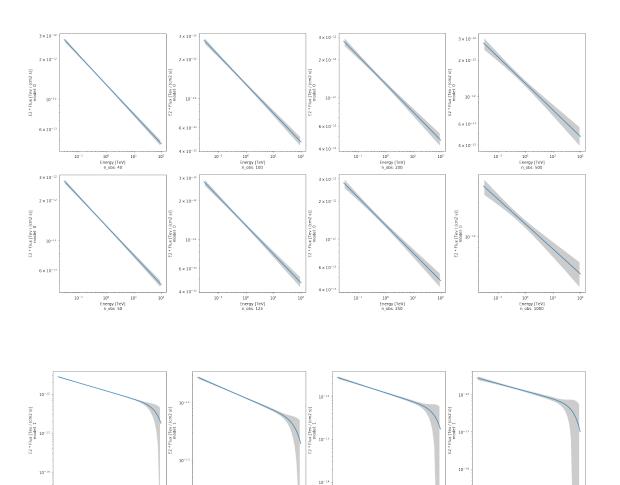
True

nan nan

alpha 2.000e+00

ExpCutoffPowerLawSpectralModel

```
ax = fig.add_subplot(gs1[n])
       sim_1[n][i].plot(energy_range=energy_range, energy_power=2)
       plot_error(self=sim_1[n][i], covar=np.mean(covar_1[n][i], axis =__
→0),energy_range=energy_range, energy_power=2)
       plt.xlabel(f'Energy [TeV]\nn_obs: {n_obs[n]}')
       plt.ylabel(f'E2 * Flux [Tev / (cm2 s)]\nmodel: {i}')
   gs2 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[1])
   for n in range(2):
       ax = fig.add_subplot(gs2[n])
       sim_1[n+2][i].plot(energy_range=energy_range, energy_power=2)
       plot_error(self=sim_1[n+2][i], covar=np.mean(covar_1[n+2][i], axis =__
→0),energy_range=energy_range, energy_power=2)
       plt.xlabel(f'Energy [TeV]\nn_obs: {n_obs[n+2]}')
       plt.ylabel(f'E2 * Flux [Tev / (cm2 s)]\nmodel: {i}')
   gs3 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[2])
   for n in range(2):
       ax = fig.add_subplot(gs3[n])
       sim_1[n+4][i].plot(energy_range=energy_range, energy_power=2)
       plot_error(self=sim_1[n+4][i], covar=np.mean(covar_1[n+4][i], axis =__
→0),energy_range=energy_range, energy_power=2)
       plt.xlabel(f'Energy [TeV]\nn obs: {n obs[n+4]}')
       plt.ylabel(f'E2 * Flux [Tev / (cm2 s)]\nmodel: {i}')
   gs4 = gridspec.GridSpecFromSubplotSpec(2, 1, subplot_spec=gs0[3])
   for n in range(2):
       ax = fig.add_subplot(gs4[n])
       sim_1[n+6][i].plot(energy_range=energy_range, energy_power=2)
       plot_error(self=sim_1[n+6][i], covar=np.mean(covar_1[n+6][i], axis =__
→0),energy_range=energy_range, energy_power=2)
       plt.xlabel(f'Energy [TeV]\nn_obs: {n_obs[n+6]}')
       plt.ylabel(f'E2 * Flux [Tev / (cm2 s)]\nmodel: {i}')
   plt.show()
```



10° Energy [TeV] n_obs: 200

10⁰ Energy [TeV] n_obs: 250 10° Energy [TeV] n_obs: 500

10° Energy [TeV] n_obs: 1000

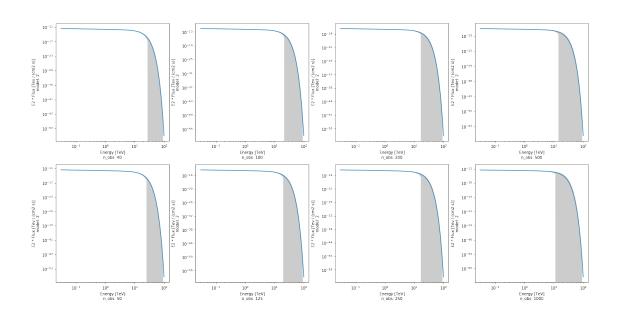
10° Energy [TeV] n_obs: 100

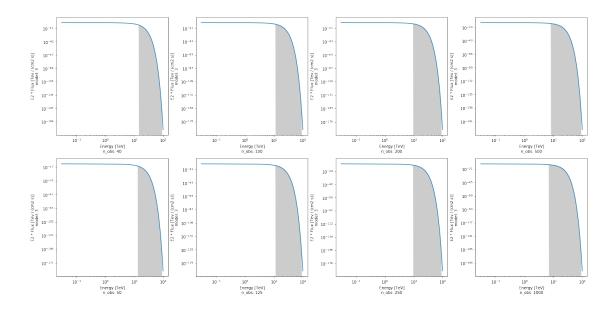
10° Energy [TeV] n_obs: 125

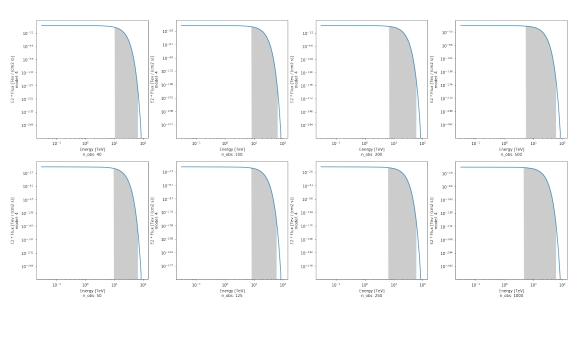
10⁰ Energy [TeV] n_obs: 40

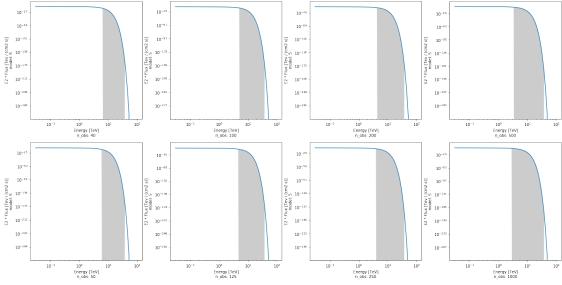
10⁰ Energy [TeV] n_obs: 50

E2 * Flux [Tev / (cm2 s)] model: 1









[]: