Example_code

February 28, 2020

1 Interpolation function needed for model

```
[27]: #Interpolation function used in model to obtain comparison points with data
def interpolate(x0, y0, x):
    x = np.array(x)
    idx = np.searchsorted(x0, x)
    dl = np.array(x - x0[idx - 1])
    dr = np.array(x0[idx] - x)
    d = dl + dr
    wl = dr / d
    return wl * y0[idx - 1] + (1 - wl) * y0[idx]
```

2 Specify simple model

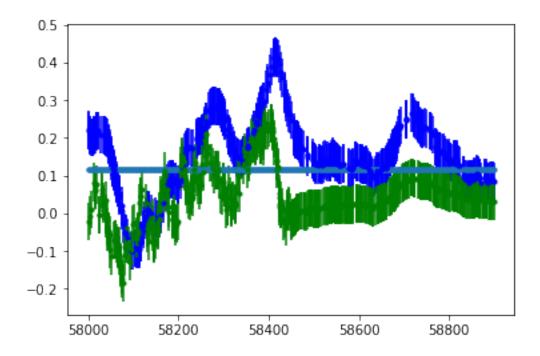
```
[48]: import random
#Load in synthetic data
yK,yKerr = np.loadtxt('Kband.txt', delimiter=',', usecols=(0,1), unpack=True)
yi,yierr = np.loadtxt('iband.txt', delimiter=',', usecols=(0,1), unpack=True)
yg,ygerr = np.loadtxt('gband.txt', delimiter=',', usecols=(0,1), unpack=True)
```

```
X=np.linspace(58000,58900,5000)
yKerr=np.ones(len(yKerr))*max(yKerr)
yierr=np.ones(len(yierr))*max(yierr)
#Select points at random
n_list = 499# Number of selected points
the_list = list(range(len(X)))
ind = random.sample(the_list, n_list)
ind2 = random.sample(the_list, n_list)
yK=yK[ind]
yKerr=yKerr[ind]
XK=X[ind]
yi=yi[ind2]
yierr=yierr[ind2]
Xi=X[ind2]
```

```
[49]: nf = 499 #Number of GP points
Xf = np.linspace(np.min([XK,Xi])-1.0, np.max([XK,Xi])+1.0,nf)
Xf=np.reshape(Xf,(len(Xf),1))
tau=np.linspace(1.0,100.0,nf)
```

```
[50]: plt.plot(Xf,2e-6*Xf,'.')
plt.errorbar(XK, yK, yKerr, fmt='b.', label='K')
plt.errorbar(Xi, yi, yierr, fmt='g.', label='i')
```

[50]: <ErrorbarContainer object of 3 artists>



```
[51]: #Define prior range
     mlow=2.3
     mhigh=3.7
     mmu=(mhigh+mlow)/2.0
     msig=(mhigh-mlow)/2.0
     siglow=0.1
     sighigh=2.55
     sigmu=(sighigh+siglow)/2.0
     sigsig=(sighigh-siglow)/2.0
[28]: with pm.Model() as convmodel:
          #define driving function as Gaussian Process
          = pm.Bound(pm.Normal, lower=0.0,upper=0.71)('', mu=0.355, sigma=0.
      →355)#timescale of variation for the driving function, order of days for UV
          #REMEMBER time scale is 2* ~2 so remember to rewrite as _true=2* ~2
           = pm.Bound(pm.Normal, lower=0.0,upper=0.4)('', mu=0.2, sigma=0.2)#lonqu
      →term standard deviation for the driving function
         cov = **2 * pm.gp.cov.Exponential(1, ls=) #using same cov as light curve_
      \rightarrow interpolation
         gp = pm.gp.Latent(cov_func=cov)
         f = gp.prior("f", X=Xf,reparameterize = False)
         f = f.reshape((1,1,len(Xf),1))
         #############
         #Define priors
         #############
         #Universal Dusty Torus parameters for the uniform temperature DT
          sigma DT=pm.Bound(pm.Normal, lower=siglow,upper=sighigh)('sigma DT', u
      →mu=sigmu, sigma=sigsig) #needs a source for scale
         m DT=pm.Bound(pm.Normal, lower=mlow,upper=mhigh)('m DT', mu=mmu,
      ⇒sigma=msig) #we expect serveral tens to hundreds of days from the nature
      \rightarrow letter
         theta_DT=pm.Bound(pm.Normal, lower=2.3,upper=4.1)('theta DT', mu=3.
      \rightarrow2, sigma=0.9) #add later when simple model is staple
          #Accretion Disk paramters
         \#K band
         Ksigma_AD=pm.Bound(pm.Normal, lower=siglow,upper=sighigh)('Ksigma_AD',__
      →mu=sigmu, sigma=sigsig) # Shappee 2014 suggests somewhere between 0-20 days
      \hookrightarrow so log that
         Ktheta AD=0.0#pm.Normal('Ktheta AD', mu=0.0, siqma=10.0)#add later
         Km_AD=pm.Bound(pm.Normal, lower=mlow,upper=mhigh)('Km_AD', mu=mmu,_
      ⇒sigma=msig) #AD has 3-5 times smaller lags than DT
          #i band
```

```
isigma_AD=pm.Bound(pm.Normal, lower=siglow,upper=sighigh)('isigma_AD',__
→mu=sigmu, sigma=sigsig) # Shappee 2014 suggests somewhere between 0-20 days
\rightarrowso log that
   itheta AD=0.0#pm.Normal('qtheta AD', mu=0.0, sigma=10.0)#add later
   im_AD=pm.Bound(pm.Normal, lower=mlow,upper=mhigh)('im_AD', mu=mmu,_
⇒sigma=msig) #AD has 3-5 times smaller lags than DT
   #BB and power law parameters
   T=pm.Bound(pm.Normal, lower=6.90775527898 ,upper=7.74066440192)('T',mu=7.
→32420984045,sigma=0.41645456146)#taken from nature letter
   K_0=pm.Bound(pm.Normal, lower=-2.3,upper=2.3)('K_0', mu=0, sigma=2.3)#powr/
\hookrightarrow BB
   index=pm.Bound(pm.Normal, lower=0.5,upper=3.0)('index', mu=1.5, sigma=0.
\hookrightarrow5)#sign depends on diffmag definition change to -2 to -1
   #Note for index: we have taken the transformation from F_nu to F_lamb into
\rightarrowaccount with the index value.
   ############
   #Define model
   ############
   #Peak Black Body from uniform torus temperature
   BB max = -79.3575 + 5.0*T#the log of the peak wavelength blackbody
   #Universal lognormal for Dusty Torus
   \exp_DT = -((tt.log((tau-tt.exp(theta_DT))/tt.exp(m_DT)))**2/(2*sigma_DT**2))
   front_DT = 1.0/((tau-tt.exp(theta_DT))*sigma_DT*tt.sqrt(2*np.pi))
   lognorm_DT = front_DT*tt.exp(exp_DT)
   lognorm_DT = tt.switch(tt.isnan(lognorm_DT), 0.0, lognorm_DT)
   #Dusty Torus transfer equation for K band
   Kb = 6692.0/tt.exp(T)
   KBB = -38.36611560560854 - tt.log(tt.exp(Kb) - 1.0)
   KPsi_DT = (tt.exp(KBB)/tt.exp(BB_max))*lognorm_DT
   #Dusty Torus transfer equation for i band
   ib = 18690.0/tt.exp(T)
   iBB = -33.230653704248226-tt.log(tt.exp(ib) - 1.0)
   iPsi_DT = (tt.exp(iBB)/tt.exp(BB_max))*lognorm_DT
   #Accretion Disk transfer equation for the K band
   Kpowr = K_0+0.649998592333457*index
   Kexp\_AD = -((tt.log((tau-Ktheta\_AD)/tt.exp(Km\_AD)))**2/(2*Ksigma\_AD**2))
   Kfront_AD = 1.0/((tau-Ktheta_AD)*Ksigma_AD*tt.sqrt(2*np.pi))
   Klognorm AD = Kfront AD*tt.exp(Kexp AD)
   KPsi_AD = tt.exp(Kpowr)*Klognorm_AD
   #Accretion Disk transfer equation for the i band
   ipowr = K_0-0.3770937879386054*index
```

```
iexp\_AD = -((tt.log((tau-itheta\_AD)/tt.exp(im\_AD)))**2/(2*isigma\_AD**2))
  ifront_AD = 1.0/((tau-itheta_AD)*isigma_AD*tt.sqrt(2*np.pi))
  ilognorm_AD = ifront_AD*tt.exp(iexp_AD)
  iPsi_AD = tt.exp(ipowr)*ilognorm_AD
  #Full transfer equations
  ##########################
  Ktransfer = KPsi DT + KPsi AD
  Ktransfer = Ktransfer.reshape(((1,1,len(tau),1)))
  itransfer = iPsi_DT + iPsi_AD
  itransfer = itransfer.reshape(((1,1,len(tau),1)))
  #The convolutions
  #'half': pad input with a symmetric border of filter rows // 2
  #rows and filter columns // 2 columns, then perform a valid convolution.
  #For filters with an odd number of rows and columns,
  #this leads to the output shape being equal to the input shape.
  Kconvol=theano.tensor.nnet.conv2d(f,Ktransfer,border_mode='half')
  Kcomp=interpolate(Xf[:,0],Kconvol[0,0,:,0],XK)
  iconvol=theano.tensor.nnet.conv2d(f,itransfer,border mode='half')
  icomp=interpolate(Xf[:,0],iconvol[0,0,:,0],Xi)
  ###################
  #Define likelihoods
  ###################
  Klikelihood = pm.Normal('yK', mu=Kcomp, sigma=yKerr, observed=yK)
  ilikelihood = pm.Normal('yi', mu=icomp, sigma=yierr, observed=yi)
  #the shape of mu and observed needs to be the same
  #max treedepth, default=10
  #The maximum tree depth. Trajectories are stopped when this depth is_{\sqcup}
\rightarrow reached.
  #early_max_treedepth, default=8
  #The maximum tree depth during the first 200 tuning samples.
  tracetransfer = pm.sample(2000,tune=2000,init='advi+adapt_diag',chains=2)
```

Auto-assigning NUTS sampler...
Initializing NUTS using advi+adapt_diag...

0%1 | 0/200000 [00:00<?, ?it/s]/home/malte/anaconda3/lib/python3.7/sitepackages/theano/tensor/basic.py:6611: FutureWarning: Using a non-tuple sequence for multidimensional indexing is deprecated; use `arr[tuple(seq)]` instead of `arr[seq]`. In the future this will be interpreted as an array index, `arr[np.array(seq)]`, which will result either in an error or a different result. result[diagonal_slice] = x Average Loss = -227.7: 12%| | 23595/200000 [02:49<21:08, 139.06it/s] Convergence achieved at 23600 Interrupted at 23,599 [11%]: Average Loss = 427.51 /home/malte/anaconda3/lib/python3.7/site-packages/theano/tensor/basic.py:6611: FutureWarning: Using a non-tuple sequence for multidimensional indexing is deprecated; use `arr[tuple(seq)]` instead of `arr[seq]`. In the future this will be interpreted as an array index, `arr[np.array(seq)]`, which will result either in an error or a different result. result[diagonal_slice] = x /home/malte/anaconda3/lib/python3.7/site-packages/theano/tensor/basic.py:6611: FutureWarning: Using a non-tuple sequence for multidimensional indexing is deprecated; use `arr[tuple(seq)]` instead of `arr[seq]`. In the future this will be interpreted as an array index, `arr[np.array(seq)]`, which will result either in an error or a different result. result[diagonal_slice] = x /home/malte/anaconda3/lib/python3.7/site-packages/theano/tensor/basic.py:6611: FutureWarning: Using a non-tuple sequence for multidimensional indexing is deprecated; use `arr[tuple(seq)]` instead of `arr[seq]`. In the future this will be interpreted as an array index, `arr[np.array(seq)]`, which will result either in an error or a different result. result[diagonal slice] = x /home/malte/anaconda3/lib/python3.7/site-packages/theano/tensor/basic.py:6611: FutureWarning: Using a non-tuple sequence for multidimensional indexing is deprecated; use `arr[tuple(seq)]` instead of `arr[seq]`. In the future this will be interpreted as an array index, `arr[np.array(seq)]`, which will result either in an error or a different result. result[diagonal slice] = x Multiprocess sampling (2 chains in 4 jobs) NUTS: [index, K_0, T, im_AD, isigma_AD, Km_AD, Ksigma_AD, theta_DT, m_DT, sigma_DT, f, ,] Sampling 2 chains, 0 divergences: 100% | 8000/8000 [12:54:30<00:00, 5.81s/draws] The acceptance probability does not match the target. It is 0.8935700370159257, but should be close to 0.8. Try to increase the number of tuning steps. The chain reached the maximum tree depth. Increase max_treedepth, increase target_accept or reparameterize. The chain reached the maximum tree depth. Increase max_treedepth, increase target_accept or reparameterize. The rhat statistic is larger than 1.4 for some parameters. The sampler did not

converge.

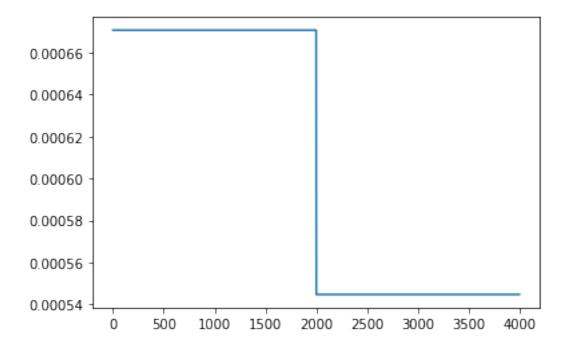
The estimated number of effective samples is smaller than 200 for some parameters.

```
[29]: tracesimple=tracetransfer
        simplenames=['sigma_DT', 'm_DT', 'theta_DT', 'isigma_AD', 'itheta_AD', 'im_AD', 'Ksigma_AD', 'Ktheta_
        simplenames=['sigma_DT','m_DT','theta_DT','isigma_AD','im_AD','Ksigma_AD','Km_AD','T','K_O','i
         \hookrightarrow '', '']
        pm.plot_posterior(tracesimple, simplenames);
                               mean=0.46
                                                                                                        mean=3.3
                                                         mean=3
                            94% HPD
                                                          94% HPD
                                                                                                     94% HPD
                                          0.53
               0.30
                     0.35
                          isigma AD
                                                               im_AD
                                                                                                Ksigma_AD
                                                                                                   94% HPD
                                 94% HPD
                                                        94% HPD
                                                                                        0.92
                           Km_AD
                                mean=2.9
                                                                                                     mean=0.31
                             94% HPD
                                                                 94% HPD
                                                                                                    4% HPD
                                           3.1
                                                                                         0.038
                                                                                                                0.56
                                                                                          0.0
                                                            7.50
                                                                                       -0.1
                                                                                              0.1
                                                                                                  0.2
                                                                                                      0.3
                             2.8
                                                              94% HPD
                                                                                             94% HPD
```

3 Sampler convergence statistics

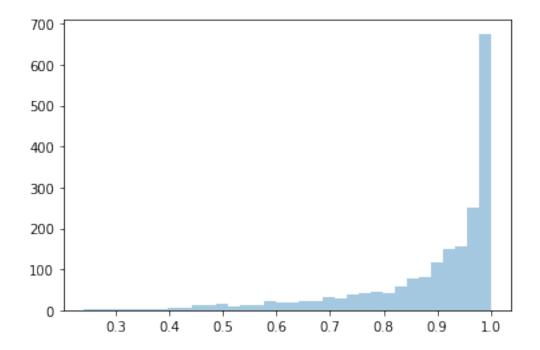
```
[30]: plt.plot(tracesimple['step_size_bar'])
```

[30]: [<matplotlib.lines.Line2D at 0x7fbb41d83f90>]



```
[31]: import seaborn as sb
accept = tracesimple.get_sampler_stats('mean_tree_accept', burn=1000)
sb.distplot(accept, kde=False)
```

[31]: <matplotlib.axes._subplots.AxesSubplot at 0x7fbb4155e6d0>

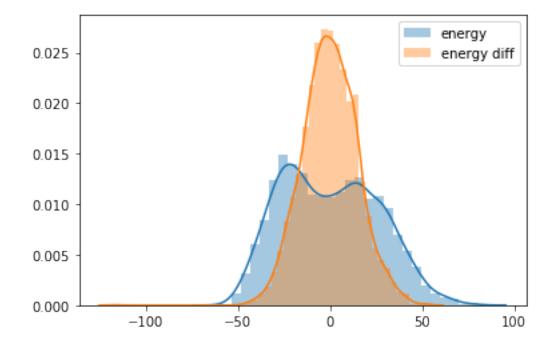


```
[56]: tracesimple['diverging'].nonzero()

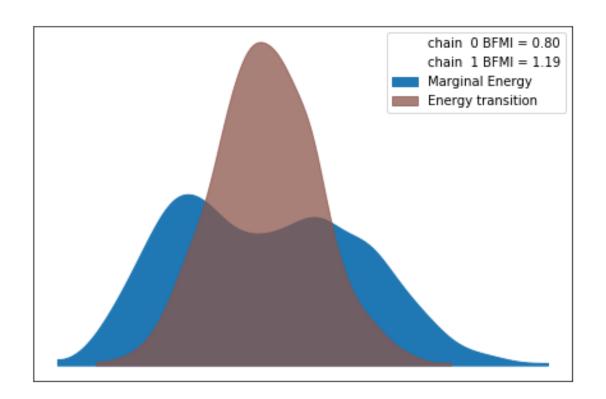
[56]: (array([], dtype=int64),)

[32]: energy = tracesimple['energy']
    energy_diff = np.diff(energy)
    sb.distplot(energy - energy.mean(), label='energy')
    sb.distplot(energy_diff, label='energy diff')
    plt.legend()
```

[32]: <matplotlib.legend.Legend at 0x7fbb424ea150>



[33]: pm.energyplot(tracesimple);



4 Sampler results

```
[34]: summ=pm.summary(tracesimple,simplenames)
      #print(summ.loc[:,'mean'])
      summ
[34]:
                            sd hpd_3% hpd_97%
                                                 mcse_mean
                                                             mcse sd
                                                                      ess mean \
                  mean
                                 0.313
                                          0.534
                                                      0.050
                                                               0.042
                                                                            2.0
      sigma_DT
                 0.457
                        0.074
      m_DT
                                                               0.092
                                                                            2.0
                 2.994
                        0.166
                                 2.820
                                          3.253
                                                      0.112
      theta_DT
                 3.296
                        0.128
                                 3.085
                                          3.410
                                                      0.087
                                                               0.073
                                                                            2.0
      isigma_AD
                 2.254
                        0.205
                                 1.862
                                          2.441
                                                      0.118
                                                               0.096
                                                                            3.0
                                          2.827
                 2.589
                                                               0.020
                                                                           32.0
      im\_AD
                        0.155
                                 2.311
                                                      0.027
      Ksigma_AD
                 1.157
                                 0.919
                                          1.362
                                                      0.107
                                                               0.090
                                                                            2.0
                        0.158
      Km_AD
                 2.915
                        0.190
                                 2.585
                                          3.108
                                                      0.121
                                                               0.100
                                                                            2.0
      Т
                 7.566
                        0.035
                                 7.490
                                          7.602
                                                      0.017
                                                               0.013
                                                                            4.0
      K_0
                 0.310
                        0.208
                                -0.038
                                          0.563
                                                      0.137
                                                               0.120
                                                                            2.0
      index
                 0.837
                        0.118
                                 0.654
                                          1.061
                                                      0.067
                                                               0.053
                                                                            3.0
                 0.314
                        0.038
                                0.272
                                          0.375
                                                     0.025
                                                               0.021
                                                                           2.0
                 0.124 0.039
                                0.082
                                          0.185
                                                     0.027
                                                               0.022
                                                                           2.0
                 ess_sd
                         ess_bulk
                                    ess_tail r_hat
                                               2.22
                    2.0
      sigma_DT
                               3.0
                                        12.0
```

m_DT	2.0	2.0	11.0	2.47
theta_DT	2.0	2.0	19.0	2.78
isigma_AD	3.0	3.0	11.0	2.29
im_AD	32.0	29.0	76.0	1.29
Ksigma_AD	2.0	3.0	14.0	2.23
Km_AD	2.0	3.0	21.0	1.85
T	4.0	4.0	55.0	1.85
K_O	2.0	3.0	12.0	2.02
index	3.0	3.0	11.0	1.62
	2.0	3.0	35.0	1.86
	2.0	2.0	11.0	2.78

[35]: pm.traceplot(tracesimple, var_names=simplenames);

/home/malte/.local/lib/python3.7/site-

packages/arviz/plots/backends/matplotlib/distplot.py:38: UserWarning: Argument backend_kwargs has not effect in matplotlib.plot_distSupplied value won't be used

"Argument backend_kwargs has not effect in matplotlib.plot_dist" /home/malte/.local/lib/python3.7/site-

packages/arviz/plots/backends/matplotlib/distplot.py:38: UserWarning: Argument backend_kwargs has not effect in matplotlib.plot_distSupplied value won't be used

"Argument backend_kwargs has not effect in matplotlib.plot_dist" /home/malte/.local/lib/python3.7/site-

packages/arviz/plots/backends/matplotlib/distplot.py:38: UserWarning: Argument backend_kwargs has not effect in matplotlib.plot_distSupplied value won't be used

"Argument backend_kwargs has not effect in matplotlib.plot_dist" /home/malte/.local/lib/python3.7/site-

packages/arviz/plots/backends/matplotlib/distplot.py:38: UserWarning: Argument backend_kwargs has not effect in matplotlib.plot_distSupplied value won't be used

"Argument backend_kwargs has not effect in matplotlib.plot_dist" /home/malte/.local/lib/python3.7/site-

packages/arviz/plots/backends/matplotlib/distplot.py:38: UserWarning: Argument backend_kwargs has not effect in matplotlib.plot_distSupplied value won't be used

"Argument backend_kwargs has not effect in matplotlib.plot_dist" /home/malte/.local/lib/python3.7/site-

packages/arviz/plots/backends/matplotlib/distplot.py:38: UserWarning: Argument backend_kwargs has not effect in matplotlib.plot_distSupplied value won't be used

"Argument backend_kwargs has not effect in matplotlib.plot_dist" /home/malte/.local/lib/python3.7/site-

packages/arviz/plots/backends/matplotlib/distplot.py:38: UserWarning: Argument backend_kwargs has not effect in matplotlib.plot_distSupplied value won't be

used "Argument backend_kwargs has not effect in matplotlib.plot_dist" /home/malte/.local/lib/python3.7/sitepackages/arviz/plots/backends/matplotlib/distplot.py:38: UserWarning: Argument backend kwargs has not effect in matplotlib.plot distSupplied value won't be "Argument backend kwargs has not effect in matplotlib.plot dist" /home/malte/.local/lib/python3.7/sitepackages/arviz/plots/backends/matplotlib/distplot.py:38: UserWarning: Argument backend_kwargs has not effect in matplotlib.plot_distSupplied value won't be used "Argument backend_kwargs has not effect in matplotlib.plot_dist" /home/malte/.local/lib/python3.7/sitepackages/arviz/plots/backends/matplotlib/distplot.py:38: UserWarning: Argument backend_kwargs has not effect in matplotlib.plot_distSupplied value won't be used "Argument backend_kwargs has not effect in matplotlib.plot_dist" /home/malte/.local/lib/python3.7/sitepackages/arviz/plots/backends/matplotlib/distplot.py:38: UserWarning: Argument backend kwargs has not effect in matplotlib.plot distSupplied value won't be used "Argument backend kwargs has not effect in matplotlib.plot dist" /home/malte/.local/lib/python3.7/sitepackages/arviz/plots/backends/matplotlib/distplot.py:38: UserWarning: Argument backend_kwargs has not effect in matplotlib.plot_distSupplied value won't be used "Argument backend_kwargs has not effect in matplotlib.plot_dist" /home/malte/.local/lib/python3.7/sitepackages/arviz/plots/backends/matplotlib/distplot.py:38: UserWarning: Argument backend_kwargs has not effect in matplotlib.plot_distSupplied value won't be used "Argument backend_kwargs has not effect in matplotlib.plot_dist" /home/malte/.local/lib/python3.7/sitepackages/arviz/plots/backends/matplotlib/distplot.py:38: UserWarning: Argument backend kwargs has not effect in matplotlib.plot distSupplied value won't be used "Argument backend_kwargs has not effect in matplotlib.plot_dist" /home/malte/.local/lib/python3.7/sitepackages/arviz/plots/backends/matplotlib/distplot.py:38: UserWarning: Argument backend_kwargs has not effect in matplotlib.plot_distSupplied value won't be used "Argument backend_kwargs has not effect in matplotlib.plot_dist" /home/malte/.local/lib/python3.7/sitepackages/arviz/plots/backends/matplotlib/distplot.py:38: UserWarning: Argument backend_kwargs has not effect in matplotlib.plot_distSupplied value won't be "Argument backend_kwargs has not effect in matplotlib.plot_dist"

/home/malte/.local/lib/python3.7/site-

packages/arviz/plots/backends/matplotlib/distplot.py:38: UserWarning: Argument backend_kwargs has not effect in matplotlib.plot_distSupplied value won't be used

"Argument backend_kwargs has not effect in matplotlib.plot_dist" /home/malte/.local/lib/python3.7/site-

packages/arviz/plots/backends/matplotlib/distplot.py:38: UserWarning: Argument backend_kwargs has not effect in matplotlib.plot_distSupplied value won't be used

"Argument backend_kwargs has not effect in matplotlib.plot_dist" /home/malte/.local/lib/python3.7/site-

packages/arviz/plots/backends/matplotlib/distplot.py:38: UserWarning: Argument backend_kwargs has not effect in matplotlib.plot_distSupplied value won't be used

"Argument backend_kwargs has not effect in matplotlib.plot_dist" /home/malte/.local/lib/python3.7/site-

packages/arviz/plots/backends/matplotlib/distplot.py:38: UserWarning: Argument backend_kwargs has not effect in matplotlib.plot_distSupplied value won't be used

"Argument backend_kwargs has not effect in matplotlib.plot_dist" /home/malte/.local/lib/python3.7/site-

packages/arviz/plots/backends/matplotlib/distplot.py:38: UserWarning: Argument backend_kwargs has not effect in matplotlib.plot_distSupplied value won't be used

"Argument backend_kwargs has not effect in matplotlib.plot_dist" /home/malte/.local/lib/python3.7/site-

packages/arviz/plots/backends/matplotlib/distplot.py:38: UserWarning: Argument backend_kwargs has not effect in matplotlib.plot_distSupplied value won't be used

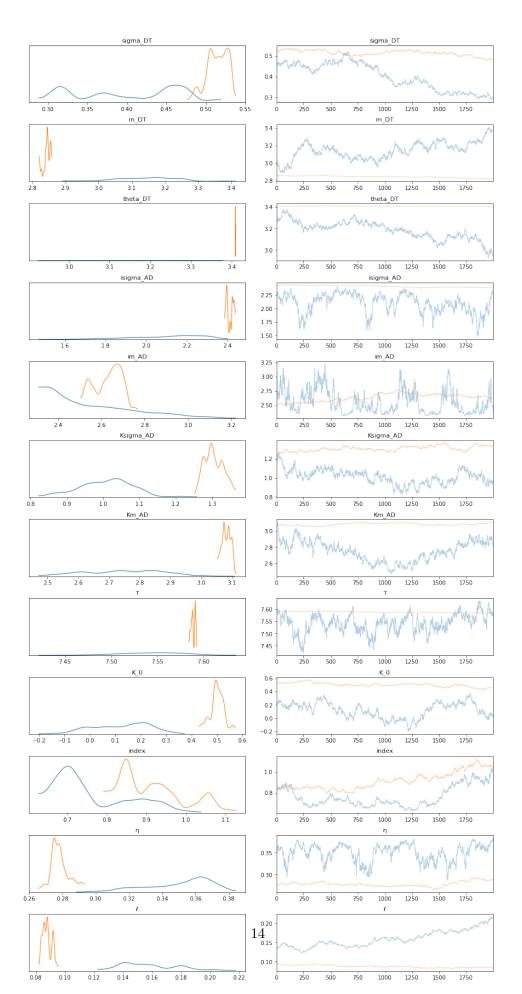
"Argument backend_kwargs has not effect in matplotlib.plot_dist" /home/malte/.local/lib/python3.7/site-

packages/arviz/plots/backends/matplotlib/distplot.py:38: UserWarning: Argument backend_kwargs has not effect in matplotlib.plot_distSupplied value won't be used

"Argument backend_kwargs has not effect in matplotlib.plot_dist" /home/malte/.local/lib/python3.7/site-

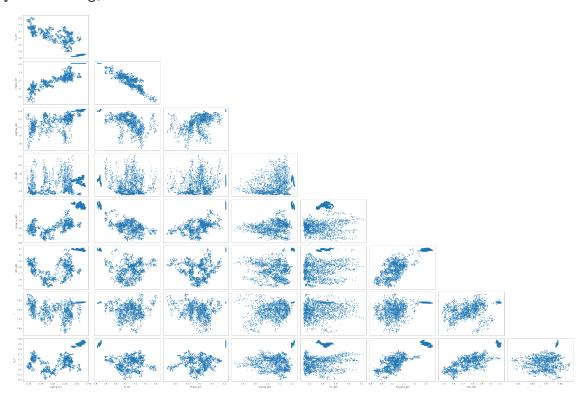
packages/arviz/plots/backends/matplotlib/distplot.py:38: UserWarning: Argument backend_kwargs has not effect in matplotlib.plot_distSupplied value won't be used

"Argument backend_kwargs has not effect in matplotlib.plot_dist"



[36]: pm.pairplot(tracesimple, var_names=simplenames, divergences=True);

/home/malte/.local/lib/python3.7/sitepackages/arviz/plots/backends/matplotlib/pairplot.py:90: SyntaxWarning: rcParams['plot.max_subplots'] (40) is smaller than the number of resulting pair plots with these variables, generating only a 9x9 grid SyntaxWarning,

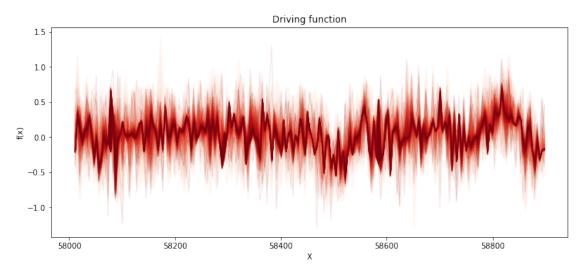


5 GP and conditional GP

```
[37]: # plot the results
fig = plt.figure(figsize=(12,5)); ax = fig.gca()

# plot the samples from the gp posterior with samples and shading
from pymc3.gp.util import plot_gp_dist
plot_gp_dist(ax, tracetransfer["f"], Xf);
#plot_gp_dist(ax, resu["f"], Xf);
# plot the data and the true latent function
#plt.plot(X, y, 'ok', ms=3, alpha=0.5, label="Observed data");
```

```
# axis labels and title
plt.xlabel("X"); plt.ylabel("f(x)");
plt.title("Driving function");
```



```
[38]: n_new = 1000
X_new = np.linspace(min(Xf), max(Xf), n_new)

# add the GP conditional to the model, given the new X values
with convmodel:
    f_J = gp.conditional("f_J", X_new)

# Sample from the GP conditional distribution
with convmodel:
    pred_samples = pm.sample_posterior_predictive(tracesimple, vars=[f_J],
    samples=1000)
```

/home/malte/.local/lib/python3.7/site-packages/pymc3/sampling.py:1247:
UserWarning: samples parameter is smaller than nchains times ndraws, some draws and/or chains may not be represented in the returned posterior predictive sample "samples parameter is smaller than nchains times ndraws, some draws "
100%| | 1000/1000 [09:58<00:00, 1.67it/s]

```
[39]: #Calculate the mean and standard deviation of the traces.
mu = np.zeros(len(X_new))
sd = np.zeros(len(X_new))

for i in range(0,len(X_new)):
    mu[i] = np.mean(pred_samples["f_J"][:,i])
    sd[i] = np.std(pred_samples["f_J"][:,i])
```

```
# draw plot
fig = plt.figure(figsize=(12,5)); ax = fig.gca()

# plot mean and 1 intervals
plt.plot(X_new, mu, 'r', lw=2, label="mean and 1 region");
plt.plot(X_new, mu + 1*sd, 'r', lw=1); plt.plot(X_new, mu - 1*sd, 'r', lw=1);
plt.fill_between(X_new.flatten(), mu - 1*sd, mu + 1*sd, color="r", alpha=0.5)

#plt.errorbar(XJ,mu,sd,fmt='r.',label='Driving function value at data points')

# plot original data and true function
#plt.plot(X, y, 'ok', ms=3, alpha=1.0, label="observed data")

#plt.errorbar(X, y, yerr,fmt='.',label="observed data")

plt.xlabel("t");
plt.ylabel("diffmag")
plt.title("predictive mean and 1 interval"); plt.legend();
print(np.mean(sd))
```

0.3430096292121166

