# **Atmosphere Plotter Notebook**

This notebook read the spectral datacubes generated by PolarVortice/AtmosphereGenerator.py.

The datacubes are located at PolarVortice/output/. In this script, I configured it to read only production datacubes.

- 1. Step 1: Read Spectral Monitoring Data
- 2. Step 2: Generate Synthetic Spectra Cube with Sonora

#### Step 0: Initialization

```
import astropy.constants as c
import astropy.units as u
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
import os
import re
import h5py
from astropy.convolution import convolve, Box1DKernel
from collections import defaultdict
import matplotlib.pyplot as plt
```

```
In [656...
          # function to print levels of nested dicts
          def print nested dict(d, indent=0):
              for key, value in d.items():
                  print(" " * indent + str(key) + ":")
                  if isinstance(value, dict):
                      print nested dict(value, indent + 1)
                  else:
                      print(" " * (indent + 1) + str(value))
          # lamda function to create nested dict
          nested_dict = lambda: defaultdict(nested_dict)
          # function to bring array value to one
          def bring_to_one(array):
              shift = (array.max() + array.min())/2
              return 1 + array - shift
          # itertools to generated random marker
          import itertools
          markerRandomList = itertools.cycle((',', '+', '.', 'o', '*'))
          linestyleRandomList = itertools.cycle(('--', '-', '--'))
          ### Usage
          # for n in y:
               plt.plot(x,n, marker = next(marker), linestyle='')
```

## Step 1: Read Spectral Monitoring Data Cube

#### Initialization

- Choose class of models to read.
- model\_class, model\_id, inclination, starting\_time, end\_time, frame\_numbers
- data structure:

dict:static|dynamic|nopolar| => dict:inclination\_value:'string' => dict:time\_value:'string' => numpy:gray\_array:2d\_image

# 1A) Read files into 1 dictionary contain 3 classes: Polar\_static, Polar\_dynamic, NoPolar

```
In [561...
         directory = "/Users/nguyendat/Documents/GitHub/polar vortice/PolarVortice/
         production0 list = []
         other list = []
         productionKey = 'production0'
         # modelclassKey = ['polarStatic', 'polarDynamic', 'noPolar']
         # modelclassKey = ['polarStatic', 'noPolar']
         print('Production key: "%s"'%productionKey)
         print('\nThe production datacubes includes:')
         for foldername in os.listdir(directory):
             if productionKey in foldername and 'polarDynamic' not in foldername:
                print(foldername)
                production0 list.append(directory+foldername+'/')
             else:
                other_list.append(directory+foldername+'/')
         print('\n')
         # Function to categorize file names
         def categorize_filenames(directory_list):
             # create empty, pre-nested dicts
             datacubes = nested_dict()
             metadatas = nested dict()
             for directory in directory list:
                 # Iterate over files in the directory
                 for filename in os.listdir(directory):
                    # print(filename)
                    # Check if the file is a H5 file
                    if filename.endswith(".h5"):
                        parts = re.findall("\[(.*?)\]", filename)
                        model_id = str(parts[1])
                        class_name = str(parts[0])
                        incli_value = str(parts[2])
                        t1t2FrameNo_value = parts[3] + "-" + parts[4] + "-" + parts
                        # Load numpy array from H5 file
                        with h5py.File(os.path.join(directory, filename), 'r') as :
                            data = f['dataset'][:]
                            datacubes[model_id][class_name][incli_value][t1t2Frame]
```

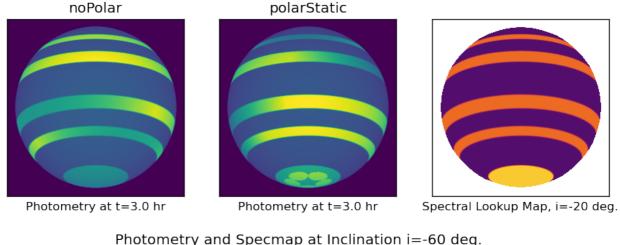
```
# Check if the file is a pickle file
          if filename.endswith(".pkl"):
             parts = re.findall("\[(.*?)\]", filename)
             model id = parts[2]
             class_name = parts[1]
             incli_value = parts[3]
             t1t2FrameNo_value = parts[4] + "-" + parts[5] + "-" + parts
             # Load metadata from pickle file
             with open(os.path.join(directory, filename), 'rb') as f:
                data = pickle.load(f)
                metadatas[model_id][class_name][incli_value][t1t2Frame]
   return datacubes, metadatas
# read the production datacube
datacubes, metadatas = categorize_filenames(production0_list)
# -----
# Print bookkeeping information
modelclasses = []
modelclasses = list(datacubes['production0'].keys())
print('=======:')
print('Datacubes contains these model classes:', modelclasses)
for model in modelclasses:
   incli = list(datacubes['production0'][model].keys())
   incli = sorted([int(x) for x in incli])
   incli = [str(x) for x in incli]
   photo config = list(datacubes['production0'][model][incli[0]].keys())
   # -----
   # Choose which photometry configuration to use:
   # Currently photo config
   photo_config(0)
   print('\n[%s] contains these inclination'%model, sorted(incli))
   print('for [t0]-[t1]-[FrameNumber]:', photo_config)
   metakeys = list(metadatas['production0'][model][incli[0]]['0-30-30'].ke
   no_frame = datacubes['production0'][model][iang][photo_config0].shape[
   t0, t1 = int(photo_config0.split('-')[0]), int(photo_config0.split('-'
   time_array = np.linspace(t0, t1, no_frame)
   print('Total no. of frames:', no frame, ', cadence: %.1f min'%((t1-t0))
print('\nThe metadata contains these attributes: \n', metakeys)
print('========')
```

```
Production key: "production0"
The production datacubes includes:
dataCube[production0][noPolar][i][0][60][60]
dataCube[production0][polarStatic][i][0][30][30]
dataCube[production0][polarStatic][i][0][60][60]
dataCube[production0][noPolar][i][0][30][30]
______
Datacubes contains these model classes: ['noPolar', 'polarStatic']
[noPolar] contains these inclination ['-10', '-20', '-30', '-40', '-50', '-
60', '-70', '-80', '-90', '0']
for [t0]-[t1]-[FrameNumber]: ['0-60-60', '0-30-30']
Total no. of frames: 60 , cadence: 60.0 min
[polarStatic] contains these inclination ['-10', '-20', '-30', '-40', '-50'
, '-60', '-70', '-80', '-90', '0']
for [t0]-[t1]-[FrameNumber]: ['0-30-30', '0-60-60']
Total no. of frames: 30 , cadence: 60.0 min
The metadata contains these attributes:
['modu_config', 'modelname', 'inclin', 'Fband', 'Fambient', 'Pband', 'Ppol
', 'config_columns', 'config', 'specmap', 'speckey', 'cond_is_amb', 'cond_i
s_band', 'cond_is_pol']
______
```

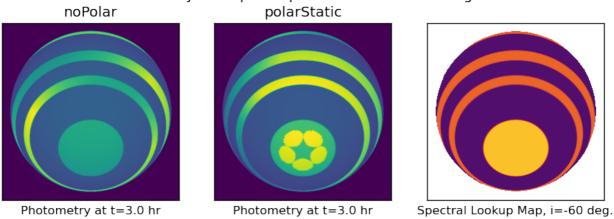
### 1B) Photometry and specmaps

```
In [633...
         # [[Plot]]
         # Plot the first photometry-frame, spectra-map for one inclin
         # of each model class
         # -----
         %matplotlib inline
         plt.close('all')
         for iang in ['-20', '-60']:
            fig, axs = plt.subplots(1, 3, dpi=120, figsize=(9,3))
            cadence = (t1-t0)/no frame
            fig.suptitle('Photometry and Specmap at Inclination i=%s deg.'%iang)
            for i, model in enumerate(modelclasses):
                photo_config(0)
                if model == 'polarStatic': timepoint = 3
                else: timepoint = 2
                gray = datacubes['production0'][model][iang][photo config0][timepo:
                axs[i].imshow(gray)
                axs[i].set title(model)
                axs[i].set_xticks([]), axs[i].set_yticks([])
                axs[i].set_xticks([]), axs[i].set_yticks([])
            specmap = metadatas['production0'][model][iang][photo config0]['specmap']
            axs[2].imshow(specmap, cmap='inferno')
            axs[2].set_xticks([]), axs[2].set_yticks([])
            axs[0].set xlabel('Photometry at t=%.1f hr'%(timepoint*cadence))
            axs[1].set xlabel('Photometry at t=%.1f hr'%(timepoint*cadence))
            axs[2].set_xlabel('Spectral Lookup Map, i=%s deg.'%iang)
```

Photometry and Specmap at Inclination i=-20 deg.



Photometry and Specmap at Inclination i=-60 deg.



## 1C) Creating flux dictionaries

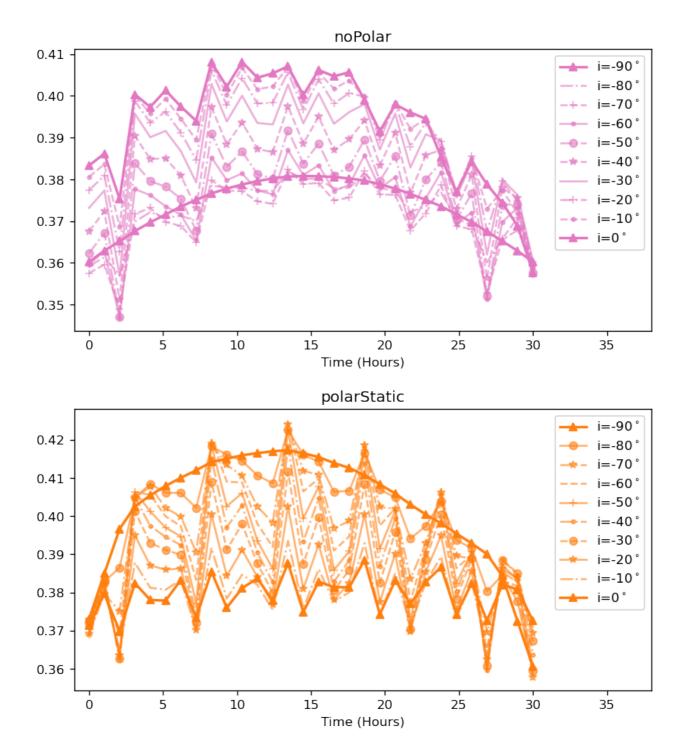
This section will assume one is using one identical time-array for 3 model classes.

```
In [661...
         # -----
         # Read photometry cubes, mask out region by feature-class
         # and save the *fluxes*, as well as *masked images*
         # and the masked specmaps *specmasks*
         # -----
         fluxes = nested_dict() # format: normflux, flux_bytype, fraction_bytype
         images = nested_dict() # format: [frameim, ambim, bandim, polim]
         specmasks = nested dict() # format: [is amb, is band, is pol]
         photo_config(0)
         do_gaussian_filter = False
         gaussian sigma = 0
         for model in modelclasses:
            for iang in incli:
                meta_iang = metadatas['production0'][model][iang][photo_config0]
                specmap = meta_iang['specmap']
                con amb = meta_iang['cond_is_amb']
                is_amb = ((specmap >= con_amb[0]) & (specmap < con_amb[1])).astype</pre>
                con_band = meta_iang['cond_is_band']
                is_band = ((specmap >= con_band[0]) & (specmap < con_band[1])).astj
                con_pol = meta_iang['cond_is_pol']
                is_pol = ((specmap >= con_pol[0]) & (specmap < con_pol[1])).astype</pre>
```

```
flux = []
fluxtyp = []
# Prepare a dictionary of specmap masks
specmap_total = metadatas['production0'][model][iang][photo_config()
specmap_amb = specmap*is_amb
specmap_band = specmap*is_band
specmap pol = specmap*is pol
specmap_total[specmap_total == 0] = np.nan
specmap amb[specmap amb == 0] = np.nan
specmap_band[specmap_band == 0] = np.nan
specmap pol[specmap pol == 0] = np.nan
frac_amb = np.nansum(specmap_amb)/np.nansum(specmap_total)
frac_band = np.nansum(specmap_band)/np.nansum(specmap_total)
frac pol = np.nansum(specmap pol)/np.nansum(specmap total)
specmasks[model][iang] = [specmap_total, specmap_amb, specmap_band]
for t in range(no frame):
        gray = datacubes['production0'][model][iang][photo config0][t]
        if do gaussian filter:
                 frameim=gaussian_filter(gray, sigma=gaussian_sigma)
        else: frameim = np.copy(gray)
        ## mask-out band and pole and calculate their
        ## respective flux contribution at each frame
        ambim, polim, bandim = frameim * is_amb, frameim * is_pol, frameim
        ambim[ambim == 0] = np.nan
        bandim[bandim == 0] = np.nan
        polim[polim == 0] = np.nan
        frameim[frameim == 0] = np.nan
        fluxtyp.append([np.nanmean(ambim), np.nanmean(bandim), np.nanme
        flux.append(np.nanmean(frameim))
        # prepare dictionary of photometries
        images[model][iang][t] = [frameim, ambim, bandim, polim]
# prepare dictionary of fluxes
flux = np.array(flux)
fluxtyp = np.array(fluxtyp)
fluxes[model][iang]['norm'] = flux
fluxes[model][iang]['bytype']['amb'] = np.transpose(fluxtyp)[0] #
fluxes[model][iang]['bytype']['band'] = np.transpose(fluxtyp)[1]
fluxes[model][iang]['bytype']['pol'] = np.transpose(fluxtyp)[2]
fluxes[model][iang]['fraction']['amb'] = frac_amb
fluxes[model][iang]['fraction']['band'] = frac_band
fluxes[model][iang]['fraction']['pol'] = frac_pol
```

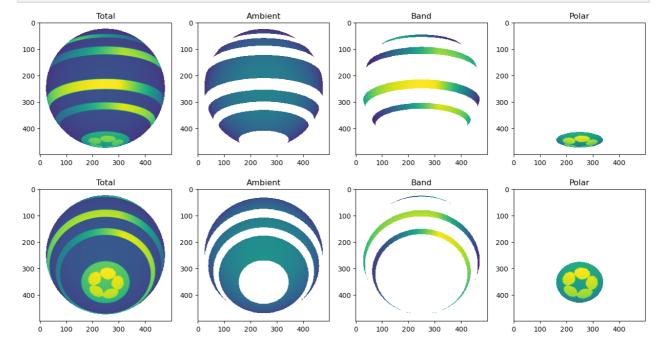
In [666...

```
# -----
# [[Plot]]
# Plot all norm flux with time for each inclinations
# in each model classes
# -----
plt.close()
colorList = {'polarStatic':'tab:orange', 'noPolar':'tab:pink', 'polarDynam'
for model in modelclasses:
   fig, ax1 = plt.subplots(dpi=120, figsize=(8,4))
   for iang in incli:
      y_flux = fluxes[model][iang]['norm']
      if iang == '-90' or iang == '0':
          line, = ax1.plot(time_array, y_flux, c=colorList[model],ls='-'
      else:
          line, = ax1.plot(time_array, y_flux, c=colorList[model],ls=nex
      ax1.set_xlim(-1, 38), ax1.set_xlabel('Time (Hours)'), ax2.set_ylabe
      ax1.set_title(model)
      ax1.legend()
```



1D) Plot photometry by each feature-class region

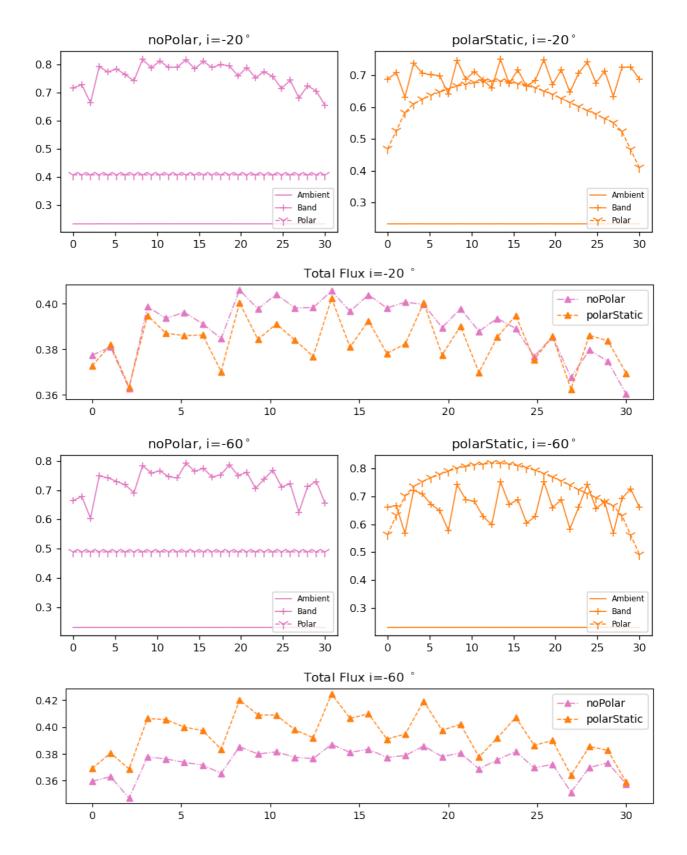
In [731...



1E) Plot flux evolution by inclination, model class, and feature class

```
In [713...
```

```
# -----
# [[Plot]]
# Plot flux by class-type (ambient, band, polar) by
# model class (polarStatic, polarDynamic, noPolar)
# for some inclination
# -----
plt.close()
colorList = {'polarStatic':'tab:orange', 'polarDynamic':'k', 'noPolar':'tal
for ia, iang in enumerate(['-20', '-60']):
   fig, axs = plt.subplots(1,2, dpi=120, figsize=(8,3))
   ticklist = {'b': '+', 'a':'', 'p':'1'}
   for i, model in enumerate(modelclasses):
       ambflux = fluxes[model][iang]['bytype']['amb']
       bandflux = fluxes[model][iang]['bytype']['band']
       polflux = fluxes[model][iang]['bytype']['pol']
       axs[i].plot(time_array, ambflux, ls='-', lw=1, c=colorList[model],
       axs[i].plot(time_array, bandflux, ls='-', lw=1, c=colorList[model]
       axs[i].plot(time array, polflux, ls='--', lw=1, c=colorList[model]
       axs[i].set_title(model+r', i=%s$^\circ$'%iang)
       axs[i].legend(fontsize=7, loc='lower right')
       plt.tight_layout()
   plt.figure(figsize=(10, 2), dpi=100)
   plt.plot(time_array, fluxes['noPolar'][iang]['norm'], ls='-.', marker=
   plt.plot(time_array, fluxes['polarStatic'][iang]['norm'], ls='--', marl
   plt.title('Total Flux i=%s $^\circ$'%iang), plt.legend()
```



Step 2: Generate Synthetic Spectra Cube with Sonora

2A) Binning down Sonora Cloudless & Cloudy Model

```
In [718...
```

```
cloudlessPath ='~/Documents/GitHub/polar_vortice/data/spectras/bobcatCloud.cloudyPath = '~/Documents/GitHub/polar_vortice/data/spectras/diamondbackCloudlessSpec = pd.read_csv(cloudlessPath, sep='\s+', names = ['wave', 'flucloudlessSpec = cloudlessSpec.query(' 0.0 <= wave <= 10') cloudlessSpec.sort_values('wave', inplace=True)

cloudySpec = pd.read_csv(cloudyPath, sep='\s+', names = ['wave', 'flux'], cloudySpec = cloudySpec.query(' 0.0 <= wave <= 10') cloudySpec.sort_values('wave', inplace=True)

cloudlessSpec['norm'] = cloudlessSpec.flux / cloudlessSpec.flux.max() cloudySpec['norm'] = cloudySpec.flux / cloudySpec.flux.max()

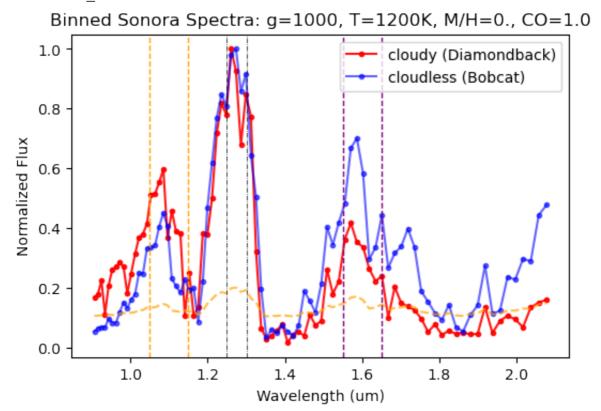
# set up color index center wavelength
lamJ, lamH = 1.10, 1.60
lamContinuum = 1.275
lamwidth = 0.05</pre>
```

#### In [855...

```
%matplotlib inline
# [[Plot]]
# Plot the binned flux, and output file for cloudy AND cloudless spectra
# idea: try searching a bin-reserving flux
# idea: try NIRSPEC resolution, R=2000
### Options
save = False ## output spectra files
# save = True
plot = False ## perform convolution of box-car smooth and plot spectras
# plot = True
read = not plot ## reading existing, smoothed spectra
# Boxcar convolution for spectra binning
if plot:
   plt.close(), plt.figure(figsize=(8,5))
   # print(cloudySpec.norm.shape)
   # plt.plot(cloudySpec.wave, cloudySpec.norm)
   binresolution = 2000
   convolve_res = 50
   wavebin1 = convolve(cloudySpec.wave, Box1DKernel(convolve res))[::binred
   normbin1 = convolve(cloudySpec.norm, Box1DKernel(convolve res))[::binred]
   # print(wavebin1.shape)
   wavebin2 = convolve(cloudlessSpec.wave, Box1DKernel(convolve_res))[::b
   normbin2 = convolve(cloudlessSpec.norm, Box1DKernel(convolve res))[::b]
   # print(wavebin2.shape)
   normbin2_interp = np.interp(x=wavebin1, xp=wavebin2, fp=normbin2)
   plt.plot(wavebin1, normbin1, ls='-', marker='.', c='r', label='cloudy
   plt.plot(wavebin1, normbin2_interp, ls='-', marker='.', c='b', label='o
   plt.plot(wavebin1, 0.1 + 0.1*(normbin2 interp+normbin1)/2, ls='--', man
   R=1.5/np.diff(wavebin1).mean()
   print('R=lamda/d lamda=', R)
```

```
# center wavelength near J and H band
   plt.axvline(lamJ-lamwidth, c='k', ls='--', lw=0.5), plt.axvline(lamJ+lamwidth, c='k', ls='--', lw=0.5),
   plt.axvline(lamH-lamwidth, c='k', ls='-.', lw=0.5), plt.axvline(lamH+lamwidth, c='k', ls='-.', lw=0.5)
   plt.axvline(lamContinuum-lamwidth/2, c='k', ls='-.', lw=0.5), plt.axvl
   plt.xlim(0.90, 2.090)
   plt.title('Binned Sonora Spectra: g=1000, T=1200K, M/H=0., CO=1.0')
   plt.xlabel('Wavelength (um)')
   plt.ylabel('Normalized Flux')
   plt.legend()
   ### Output binned spectras into text files
   pathCloudyOut = '~/Documents/GitHub/polar_vortice/data/spectras/binR=%
   cloudyOut = pd.DataFrame(np.transpose(np.array([wavebin1, normbin1])),
   cloudyOut = cloudyOut.query('0.90 <= wave <= 2.09')</pre>
   cloudyOut.flux = cloudyOut.flux/cloudyOut.flux.max()
   if save: cloudyOut.to csv(pathCloudyOut, index=False)
   pathCloudlessOut = '~/Documents/GitHub/polar vortice/data/spectras/binl
   cloudlessOut = pd.DataFrame(np.transpose(np.array([wavebin1, normbin2_:
   cloudlessOut = cloudlessOut.query('0.90 <= wave <= 2.09')</pre>
   cloudlessOut.flux = cloudlessOut.flux/cloudlessOut.flux.max()
   if save: cloudlessOut.to_csv(pathCloudlessOut, index=False)
# read from existing saved spectras
if read:
   header = ['wave', 'flux']
   pathCloudyOutput = '~/Documents/GitHub/polar vortice/data/spectras/binl
   cloudy = pd.read csv(pathCloudyOutput, names=header, skiprows=1)
   pathCloudlessOutput = '~/Documents/GitHub/polar vortice/data/spectras/l
   cloudless = pd.read_csv(pathCloudlessOutput, names=header, skiprows=1)
   plt.close(), plt.figure(figsize=(6,4), dpi=100)
   # print(cloudy.flux.shape)
   bandspec = cloudy.flux
   polarspec = cloudless.flux
   basespec = 0.1 + 0.1*cloudless.flux
   lam = cloudy.wave
   plt.plot(lam, bandspec, ls='-', marker='.', c='r', label='cloudy (Diame
   plt.plot(lam, polarspec, ls='-', marker='.', c='b', label='cloudless (1
   plt.plot(lam, basespec, ls='--', marker='', c='orange', alpha=0.7)
   R=1.5/np.diff(cloudy.wave).mean()
   print('R=lamda/d_lamda=', R)
   # center wavelength near J and H band
   plt.axvline(lamJ-lamwidth, c='orange', ls='--', lw=1), plt.axvline(lamble)
   plt.axvline(lamH-lamwidth, c='purple', ls='--', lw=1), plt.axvline(laml
   plt.axvline(lamContinuum-lamwidth/2, c='k', ls='-.', lw=0.5), plt.axvl
   plt.title('Binned Sonora Spectra: g=1000, T=1200K, M/H=0., CO=1.0')
   plt.xlabel('Wavelength (um)')
   plt.ylabel('Normalized Flux')
   plt.legend()
```

R=lamda/d\_lamda= 106.21354832727827



## 2B) Generate the spectral datacube

#### Method of generate spectral cube:

- Obtain flux by feature-class type ( ambient , band , polar ): These types will be a function of:
  - Latitudinal distribution of each feature-class
  - and the inclination angle which affect the projected area of each feature-class
- Generate spectras by feature-class type ( ambient , band , polar ):
  - Use a constant 0.2 flux for ambient .
  - Use Sonora-Bobcat Cloudless for band .
  - Use Sonora-Diamondback Cloudy for polar.
- Final expression: Total = norm\_mean(  $\Sigma$ [ Frac. area $(i) \times (1 + \operatorname{Flux}(i)) \times \operatorname{Spectra}(i)$ ] )

```
In [856...
```

```
# -----
# Define spectras of ambient (basespec),
# band (bandspec) and polar region (polarspec).
#### Output options
save = False
# save = True
testPlot = True
# testPlot = False
bandspec = cloudy.flux
polarspec = cloudless.flux
basespec = 0.1 + 0.1*cloudless.flux
lam = cloudy.wave
# outline the wavelength band to calculate color
Jband_index = np.where(np.logical_and((lamJ-lamwidth)<=lam, lam<=(lamJ+lamvidth)</pre>
Hband index = np.where(np.logical and((lamH-lamwidth) <= lam, lam <= (lamH+lamwidth) <= <= (lamH+lamwidth)
continuum index = np.where(np.logical and((lamContinuum-lamwidth/2)<=lam,</pre>
# print(lam[Jband index], lam[Hband index])
# Spectra cube configs
spectral_cube = nested_dict()
colorList = {'polarStatic':'tab:orange', 'polarDynamic':'k', 'noPolar':'tal
for i, model in enumerate(modelclasses):
       for iang in incli:
               ambflux = fluxes[model][iang]['bytype']['amb']
               bandflux = fluxes[model][iang]['bytype']['band']
               polflux = fluxes[model][iang]['bytype']['pol']
               frac_amb = fluxes[model][iang]['fraction']['amb']
               frac band = fluxes[model][iang]['fraction']['band']
               frac_pol = fluxes[model][iang]['fraction']['pol']
               # prepare spectral cube
               spectra array, JHcolor array = [], []
               continuum array = []
               for i,t in enumerate(time array):
                       spectra_at_t = ambflux[i]*frac_amb*basespec + bandflux[i]*frac
                       color at_t = np.mean(spectra_at_t[Jband_index]) - np.mean(spectra_at_t[Jband_index])
                      continuumRatio = color_at_t/np.mean(spectra_at_t[continuum_inde
                       spectra array.append(spectra at t)
                       JHcolor_array.append(color_at_t)
                       continuum array.append(continuumRatio)
               spectral_cube[model][iang]['spectra'] = spectra_array
               spectral_cube[model][iang]['JH_color'] = JHcolor_array
               spectral_cube[model][iang]['continuumRatio'] = continuum_array
```

# 2C) Spectra cube plot: all spectras through time, and J-H color by model class

```
• J color: mean(\lambda=1.05-1.15\mu m)
• H color: mean(\lambda=1.55-1.65\mu m)
• Continuum: mean(\lambda=1.250-1.300\mu m)
```

```
In [857...
```

```
# -----
# [[Plot]]
# Spectra cube test plots for two inclination, by model class
# 1) spectras at all timestamps, two model classes: blue and red
# 2) j-h color evolution by time, two model classes: blue and red
# -----
plt.close()
for iang in ['0', '-20', '-60', '-90']:
   f, axs = plt.subplots(1, 3, figsize=(11,4), dpi=100, gridspec kw={'widton'
   cadence = (t1-t0)/no frame
   f.suptitle('Spectra-time-series & J-H color, i=%s deg.'%iang)
   axs[0].set_title('Spectra at all Timestamp')
   axs[1].set_title('J-H color over time')
   offset = [0., 0.5]
   for i, model in enumerate(modelclasses):
       linestyle = next(linestyleRandomList)
       for t in range(len(time_array)):
           specnum = spectral cube[model][iang]['spectra'][t]
           axs[0].plot(lam, specnum +offset[i], ls=linestyle, lw=0.2, marl
       axs[0].plot([], c=colorList[model], label=model)
       axs[0].legend()
       axs[0].set_xlabel('Wavelength (um)')
       axs[0].set_ylabel('Intensity')
       axs[0].axhline(offset[i], ls='--', lw=0.2, c='k')
       colorArray = spectral cube[model][iang]['JH color']
       axs[1].plot(time_array, colorArray, c=colorList[model], ls='--', lv
       axs[1].legend()
       axs[0].set ylabel('[J-H color]')
       axs[0].set_xlabel('Time (hour)')
       axs[2].imshow(specmasks[model][iang][0], cmap='inferno', vmax=0.85
       axs[2].set_xticks([]), axs[2].set_yticks([])
   plt.tight layout()
```

Spectra-time-series & J-H color, i=0 deg. Spectra at all Timestamp J-H color over time noPolar noPolar 0.030 0.8 polarStatic polarStatic 0.028 0.6 [J-H color] 0.026 0.024 0.2 0.022 0.0 1.0 2.0 1.8 Time (hour) Spectra-time-series & J-H color, i=-20 deg. Spectra at all Timestamp J-H color over time 0.024 noPolar 0.8 polarStatic 0.022 0.020 0.6 [J-H color] 0.018 noPolar polarStatic 0.4 0.016 0.014 0.012 0.0 0.010 1.0 1.2 2.0 Time (hour) Spectra-time-series & J-H color, i=-60 deg. Spectra at all Timestamp J-H color over time noPolar -0.005 polarStatic 0.8 -0.010 0.6 []-H color] noPolar -0.015 polarStatio -0.020 0.2 -0.025 0.0 1.6 1.0 1.2 1.4 1.8 2.0 10 15 30 20 Time (hour) Spectra-time-series & J-H color, i=-90 deg. Spectra at all Timestamp J-H color over time -0.010 noPolar polarStatic 0.8 -0.015 0.6 []-H color] -0.020 noPolar polarStatio -0.025 0.2 -0.030 0.0 -0.035 1.2 1.8

2D) Spectra cube plot: all spectras through time, and [J-H]/continuum by model class

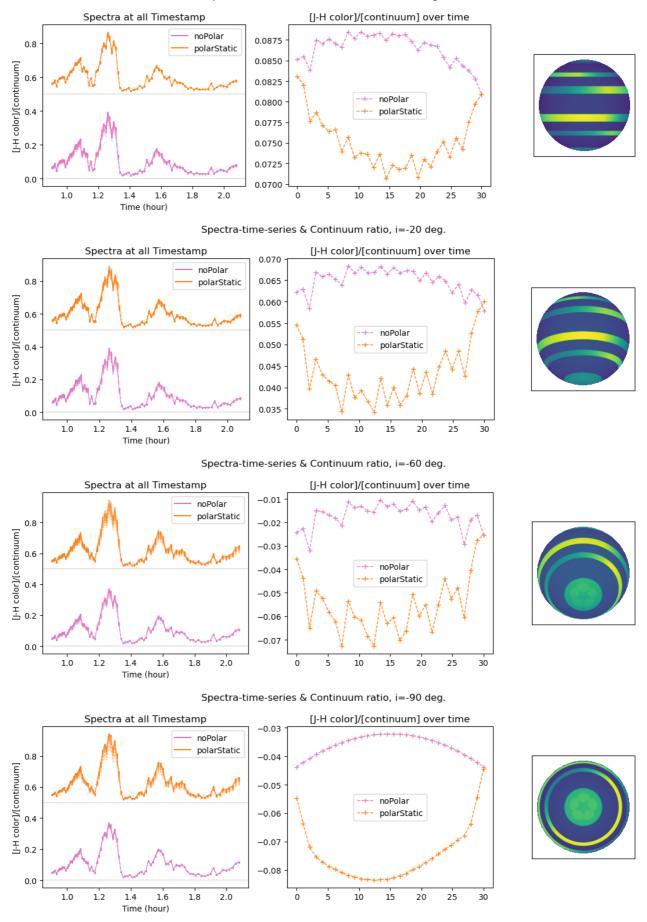
1.6

- J color:  $\mathrm{mean}(\lambda=1.05-1.15\mu m)$
- H color: mean( $\lambda = 1.55 1.65 \mu m$ )
- Continuum: mean $(\lambda = 1.250 1.300 \mu m)$

In [858...

```
# ------
# [[Plot]]
# Similar to the above plot, but use [j-h color]/[continuum] (continuum at
# Spectra cube test plots for two inclination, by model class
# 1) spectras at all timestamps, two model classes: blue and red
# 2) [j-h color]/[continuum] evolution by time, two model classes: blue an
# -----
plt.close()
for iang in ['0', '-20', '-60', '-90']:
   f, axs = plt.subplots(1, 3, figsize=(11,4), dpi=100, gridspec kw={'widton'
   cadence = (t1-t0)/no frame
   f.suptitle('Spectra-time-series & Continuum ratio, i=%s deg.'%iang)
   axs[0].set_title('Spectra at all Timestamp')
   axs[1].set_title('[J-H color]/[continuum] over time')
   offset = [0., 0.5]
   for i, model in enumerate(modelclasses):
       linestyle = next(linestyleRandomList)
       for t in range(len(time_array)):
           specnum = spectral cube[model][iang]['spectra'][t]
           axs[0].plot(lam, specnum +offset[i], ls=linestyle, lw=0.2, marl
       axs[0].plot([], c=colorList[model], label=model)
       axs[0].legend()
       axs[0].set_xlabel('Wavelength (um)')
       axs[0].set_ylabel('Intensity')
       axs[0].axhline(offset[i], ls='--', lw=0.2, c='k')
       continuumArray = spectral cube[model][iang]['continuumRatio']
       axs[1].plot(time_array, continuumArray, c=colorList[model], ls='--
       axs[1].legend()
       axs[0].set_ylabel('[J-H color]/[continuum]')
       axs[0].set_xlabel('Time (hour)')
       axs[2].imshow(images[model][iang][0][0])
       axs[2].set_xticks([]), axs[2].set_yticks([])
   plt.tight layout()
```

Spectra-time-series & Continuum ratio, i=0 deg.



2E) Plot amplitude of J-H color time-variation as function of inclination for model class

Definition of variation amplitude:  $\mathrm{Amp}(x) = (\max(x) - \min(x))/2$ 

```
In [859...
```

```
# -----
# [[Plot]]
# Quantify the variability amplitude of (1) and (2) as a function of incli
# by model class
# (1) J-H color, (2) [J-H color]/[continuum]
amplitudes = nested dict()
for model in modelclasses:
   emp1, emp2 = [], []
   # print(model)
   for iang in incli:
       continuumArray = np.array(spectral cube[model][iang]['continuumRat']
       colorArray = np.array(spectral_cube[model][iang]['JH_color'])
       ampColor = (colorArray.max()-colorArray.min())/2
       # print('i=', iang, colorArray.max(), colorArray.min(), ampColor)
       ampContinuumRatio = (continuumArray.max()-continuumArray.min())/2
       empl.append([int(iang), ampColor])
       emp2.append([int(iang), ampContinuumRatio])
   amplitudes[model]['JH_color'] = np.array(emp1)
   amplitudes[model]['continuumRatio'] = np.array(emp2)
   # print('======')
   # print(amplitudes[model]['JH color'])
# -----
## The plot: J-H color
# ------
plt.close()
titles = ['No polar variation', 'With polar variation']
# for typename in ['JH_color', 'continuumRatio']:
for typename in ['JH_color']:
   fig, axs = plt.subplots(1,4, figsize=(12,3), dpi=120, gridspec kw={'wio
   # fig.suptitle('%s with inclination i'%typename)
   for i, model in enumerate(modelclasses):
       x = amplitudes[model][typename][:,0]
       y = amplitudes[model][typename][:,1]
       axs[2*i].plot(x, y, ls='--', marker='*', c=colorList[model], ms=10
       axs[2*i].legend()
       axs[2*i].set xlabel(r'Inclination angle ($^\circ$)')
       axs[2*i].set_title(titles[i])
   axs[0].text(-90, 0.0025, 'Pole-on', fontsize=8), axs[0].text(-18, 0.002
   axs[2].text(-90, 0.006, 'Pole-on', fontsize=8), axs[2].text(-18, 0.006
   axs[0].set_ylabel('J-H color')
   axs[1].imshow(images['noPolar']['-50'][time1][0])
   axs[1].set_xticks([]), axs[1].set_yticks([])
   time2 = 3
   axs[3].imshow(images['polarStatic']['-50'][time2][0])
   axs[3].set_xticks([]), axs[3].set_yticks([])
   plt.tight layout()
```

