Jaeggli etal 2022ApJ AdHoc Xtalk

March 6, 2022

This is a Python 3 notebook is inteded to accompany Jaeggli, Schad, Tarr, & Harrington (2022, submitted) "A Model-based Technique for Ad Hoc Correction of Instrumental Polarization in Solar Spectropolarimetry" to demonstrate the application of the model-based ad hoc technique of polarization cross-talk removal.

```
[1]: # Make sure you have all of these modules installed
     import numpy as np
     from matplotlib import colors, pyplot as plt
     from astropy.io import fits
     from glob import glob
     from scipy.optimize import minimize
     from scipy.ndimage import shift
     from sklearn import linear_model
     import pandas as pd
     # Set this to your local directory with the Hinode SOT/SP raster
     datadir = './data/20100803_150053_level1/'
     # Restore the DKIST cross-talk Mueller matrix
     MMsys = np.load('./github/MMsys.npy')
     # Set some pixel ranges and thresholds
     c0 = 1 # pixel range for the continuum region
     c1 = 15
     10 = 75 # approximate center pixel of the Fe I 6302.5 line
     dw1 = 2 \# +/- pixel range for the line core
     dw2 = 14 \# +/- pixel range for the line wing
     ithresh = 0.465 # continuum intensity threshold for the sunspot umbra
     pthresh = 0.05 # polarization threshold for weak/strong polarization regions
```

```
[2]: # Read in the Hinode SOT/SP data
files = glob(datadir+'*.fits')
files = np.sort(files)

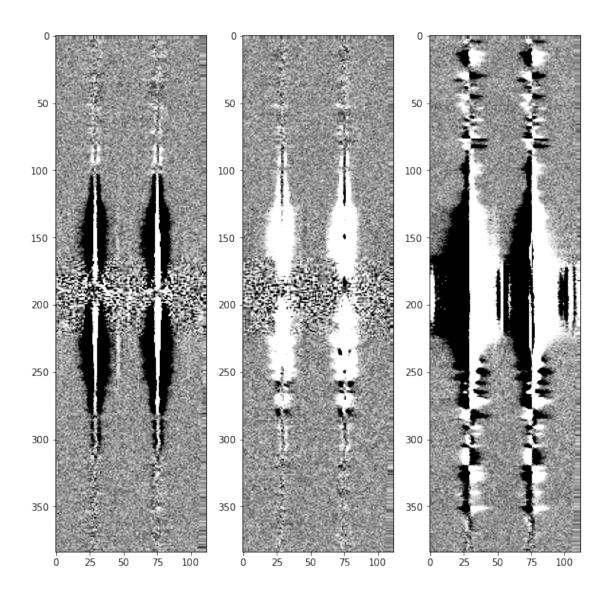
nZ = len(files) # spatial dimension perpendicular to the slit
```

```
# Open the first file and look at the header to get data dimensions
     with fits.open(files[0]) as hdu:
         idx=hdu[0].header
     nX = idx['NAXIS1'] # wavelength dimension
     nY = idx['NAXIS2'] # spatial dimension along the slit
     nS = idx['NAXIS3'] # number of Stokes vector elements
     data = np.ndarray([nZ,nY,nX,nS])
     #index = list()
     for iZ in range(0,nZ):
         with fits.open(files[iZ]) as hdu:
             dat = hdu[0].data
         data[iZ,:,:,:] = dat.transpose([1,2,0])
     # Normalize data to Stokes I
     norm = np.median(data[:,:,:,0])
     data = data/norm
     # Destreak Stokes V so any residuals from SP_PREP do not impact our work
     for iZ in range(0,nZ):
         for iY in range(0,nY):
             # Threshold Stokes V
             if np.mean(np.abs(data[iZ,iY,:,3]/data[iZ,iY,:,0])) < 0.08:</pre>
                 good=np.argwhere(np.abs(data[iZ,iY,:,3]/data[iZ,iY,:,0]) < 0.01)
                 good=good[:,0]
                 medval=np.median(data[iZ,iY,good,3]/data[iZ,iY,good,0])
                 data[iZ,iY,:,3]=data[iZ,iY,:,3]-medval*data[iZ,iY,:,0]
[3]: # Show that continuum polarization is removed from the regions that need it
     fig,axs = plt.subplots(ncols=3,nrows=1)
     fig.set_size_inches([10,10])
     iZ = 180 # region from the middle of the sunspot
```

```
3]: # Show that continuum polarization is removed from the regions that need it
fig.axs = plt.subplots(ncols=3,nrows=1)
fig.set_size_inches([10,10])

iZ = 180 # region from the middle of the sunspot

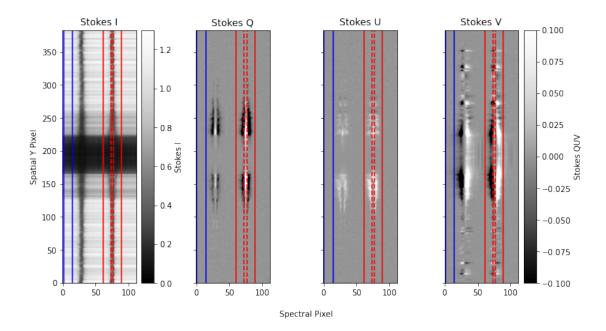
axs[0].imshow(data[iZ,:,:,1]/data[iZ,:,:,0], vmin=-0.01, vmax=0.01,
cmap='Greys_r')
axs[1].imshow(data[iZ,:,:,2]/data[iZ,:,:,0], vmin=-0.01, vmax=0.01,
cmap='Greys_r')
axs[2].imshow(data[iZ,:,:,3]/data[iZ,:,:,0], vmin=-0.01, vmax=0.01,
cmap='Greys_r')
plt.show()
```



```
axs[1].imshow(data[iZ,:,:,1], vmin=-0.1, vmax=0.1, origin='lower',_
 ⇔cmap='Greys_r')
axs[1].set_title('Stokes Q')
axs[2].imshow(data[iZ,:,:,2], vmin=-0.1, vmax=0.1, origin='lower',_
axs[2].set_title('Stokes U')
implot2=axs[3].imshow(data[iZ,:,:,3], vmin=-0.1, vmax=0.1, origin='lower',_

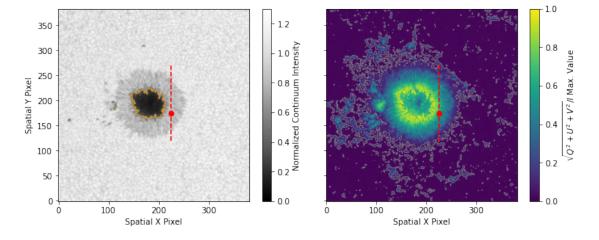
cmap='Greys_r')

axs[3].set_title('Stokes V')
cb2=fig.colorbar(implot2, ax=axs[3], label='Stokes QUV')
for iS in range(0,nS):
    #show the continuum region
   axs[iS].plot([c0,c0],[0,nY-1], 'b-')
   axs[iS].plot([c1,c1],[0,nY-1], 'b-')
    #show the line center
    \#axs[iS].plot([l0,l0],[0,nY-1], 'r--')
   #show the line core
   axs[iS].plot([10-dw1,10-dw1],[0,nY-1], 'r--')
   axs[iS].plot([10+dw1,10+dw1],[0,nY-1], 'r--')
   #show the line wings
   axs[iS].plot([10-dw2,10-dw2],[0,nY-1], 'r-')
   axs[iS].plot([10+dw2,10+dw2],[0,nY-1], 'r-')
plt.tight_layout()
plt.show()
```



```
[5]: # Do selection of spatial areas
     x0 = 10-dw2
     x1 = 10+dw2
     # Make the continuum intensity map
     imap = np.median(data[:,:,c0:c1,0], axis=2)
     imap = imap.transpose()
     # Make the polarization fraction map
     pmap = np.max( np.sqrt(np.sum(data[:,:,x0:x1,1:]**2, axis=3))/data[:,:,x0:
      \hookrightarrowx1,0], axis=2)
     pmap = pmap.transpose()
     # Apply thresholds to define different regions
     isumbra = np.argwhere(imap < ithresh)</pre>
     uyidx = isumbra[:,0]
     uzidx = isumbra[:,1]
     ispolar = np.argwhere(pmap > pthresh)
     pyidx = ispolar[:,0]
     pzidx = ispolar[:,1]
     notpolar = np.argwhere(pmap < pthresh)</pre>
     nyidx = notpolar[:,0]
     nzidx = notpolar[:,1]
```

```
[6]: %matplotlib inline
     fig,axs=plt.subplots(ncols=2, nrows=1, sharex=True, sharey=True)
     fig.set_size_inches(10,4)
     iZ = 225
     iY=175
     y0 = 120
     y1 = 270
     iplot=axs[0].imshow(imap, vmin=0.0, vmax=1.3, origin='lower', cmap='Greys_r')
     axs[0].contour(imap, [ithresh], colors='orange', linewidths=0.75)
     axs[0].set_xlabel('Spatial X Pixel')
     axs[0].set_ylabel('Spatial Y Pixel')
     icb=fig.colorbar(iplot, ax=axs[0], label='Normalized Continuum Intensity')
     pplot=axs[1].imshow(pmap, vmin=0, vmax=1.0, origin='lower')
     axs[1].contour(pmap, [pthresh], colors='gray', linewidths=0.75)
     axs[1].set_xlabel('Spatial X Pixel')
     pcb=fig.colorbar(pplot, ax=axs[1], label='\sqrt{Q^2+U^2+V^2}/I Max. Value')
     axs[0].plot([iZ,iZ], [y0,y1], '--r')
     axs[0].plot([iZ], [iY], 'or')
     axs[1].plot([iZ,iZ], [y0,y1], '--r')
     axs[1].plot([iZ], [iY], 'or')
     plt.tight_layout()
     plt.show()
```



[7]: # These are the model and minimization functions as defined in the paper
Functions for the diattenuation modeling

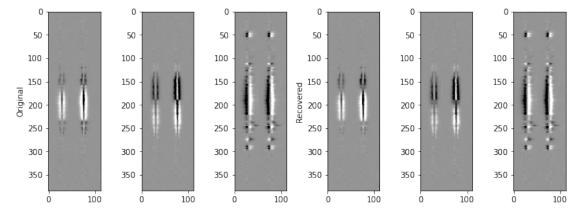
```
def polmodel1(D,theta,chi):
    dH = D*np.cos(chi)*np.sin(theta)
    d45 = D*np.sin(chi)*np.sin(theta)
    dR = D*np.cos(theta)
    A = np.sqrt(1. - dH**2 - d45**2 - dR**2)
    mat1 = np.array([
        [ 1., dH, d45, dR],
        [dH, A, O., O.],
        [d45, 0., A, 0.],
        [ dR, O., O., A]], dtype='double')
    mat2 = np.array([
        [0.,
               0.,
                       0.,
                              0.],
        [0., dH**2, d45*dH, dH*dR],
        [0., d45*dH, d45**2, d45*dR],
        [0., dH*dR, d45*dR, dR**2]], dtype='double')
    return( mat1 + (1-A)/D**2*mat2 )
def fitfunc1(param, stokesin):
   D = param[0]
    theta = param[1]
    chi = param[2]
    # Keep diattenuation value in range
    if D>=1:
       D=0.999999
    if D<=-1:
       D = -0.999999
   MM = polmodel1(D, theta, chi)
    iMM = np.linalg.inv(MM)
    out = minimize_for_model1(iMM,stokesin)
    return(out)
def minimize_for_model1(iMM,bs):
    # apply a mueller matrix (rotation) to a 2D stokes vector_{\sqcup}
 \hookrightarrow (slit_Y, wavelength_X,4)
    new_stokes = np.einsum('ij,abj->abi',iMM, np.squeeze(bs))
    # Minimization criteria
    out = np.abs(np.sum(new_stokes[:,:,0]*new_stokes[:,:,3],axis=1)) + \
          np.abs(np.sum(new_stokes[:,:,0]*new_stokes[:,:,2],axis=1)) +
```

```
np.abs(np.sum(new_stokes[:,:,0]*new_stokes[:,:,1],axis=1))
    # sum over spatial positions
    out = np.sum(out)
    return(out)
# Functions for the retarder modeling
def polmodel2(theta, delta):
    St = np.sin(theta)
    Ct = np.cos(theta)
    Sd = np.sin(delta)
    Cd = np.cos(delta)
    MM1 = np.array([
        [1., 0., 0., 0.],
        [0., Ct, St, 0.],
        [0., -St, Ct, 0.],
        [0., 0., 0., 1.]
    ], dtype='double')
    MM2 = np.array([
        [1., 0., 0., 0.],
        [0., 1., 0., 0.],
        [0., 0., Cd, Sd],
        [0., 0., -Sd, Cd]
    ], dtype='double')
    MM = np.einsum('ij,jk', MM1, MM2)
    return(MM)
def fitfunc2(fitangles, stokesin):
    theta = fitangles[0]
    delta = fitangles[1]
    MM = polmodel2(theta, delta)
    iMM = np.linalg.inv(MM)
    out = minimize_for_model2(iMM, stokesin)
    return(out)
def minimize_for_model2(iMM,bs):
    new_stokes = np.einsum('ij,abj->abi',iMM, np.squeeze(bs))
    # Minimization criteria
```

```
[8]: # Apply cross-talk Mueller matrix to the original data
     data0 = np.einsum('ij,abcj->abci', MMsys, data)
     # Choose initial quess parameters for the diattenuation minimization
     D = 0.5
     theta = 0.
     chi = 0.
     initial_guess = (D, theta, chi)
     # Use just the region with weak polarization
     baddata = data0[nzidx,nyidx,:,:] #do selection for only strong polarization_
      ⇔signals
     result = minimize(fitfunc1, initial_guess, args=baddata)
     # Apply correction for I<->QUV cross-talk
     MM1a = polmodel1(result.x[0],result.x[1], result.x[2])
     iMM1a = np.linalg.inv(MM1a)
     data1a = np.einsum('ij,abcj->abci', iMM1a, data0)
     # Destreaking correction to data might happen here
     # Choose an initial guess of those cross-talk parameters for the minimization
     ⇔algorithm
     theta = 0.*np.pi/180.
     delta = 0.*np.pi/180.
     initial_guess = (theta,delta)
     # Find the elliptical retardance parameters that minimize V*Q, V*U, and V*V
     baddata = data1a[pzidx, pyidx,:,:]
     result = minimize(fitfunc2, initial_guess, args=baddata)
     # Apply correction for QU<->V cross-talk
     MM2a = polmodel2(result.x[0],result.x[1])
     iMM2a = np.linalg.inv(MM2a)
     data2a = np.einsum('ij,abcj->abci', iMM2a, data1a)
```

```
[9]: # Apply final sign correction to match original spectra iMM3a = np.array( [[1., 0., 0., 0.],
```

```
[0., 1., 0., 0.],
                   [0., 0., 1., 0.],
                   [0., 0., 0., 1.]], dtype='double')
MM3a = np.linalg.inv(iMM3a)
data3a = np.einsum('ij,abcj->abci', iMM3a, data2a)
MMa=MM1a@MM2a@MM3a
# Show a quick plot to verify
fig,axs = plt.subplots(ncols=6)
fig.set_size_inches(10,5)
iZ = 225 # region from the middle of the sunspot
axs[0].imshow(data[iZ,:,:,1], vmin=-0.1, vmax=0.1, cmap='Greys r')
axs[1].imshow(data[iZ,:,:,2], vmin=-0.1, vmax=0.1, cmap='Greys_r')
axs[2].imshow(data[iZ,:,:,3], vmin=-0.1, vmax=0.1, cmap='Greys r')
axs[0].set_ylabel('Original')
axs[3].imshow(data3a[iZ,:,:,1], vmin=-0.1, vmax=0.1, cmap='Greys_r')
axs[4].imshow(data3a[iZ,:,:,2], vmin=-0.1, vmax=0.1, cmap='Greys_r')
axs[5].imshow(data3a[iZ,:,:,3], vmin=-0.1, vmax=0.1, cmap='Greys_r')
axs[3].set_ylabel('Recovered')
plt.tight_layout()
plt.show()
```



```
[10]: # Apply the Sanchez Almeida & Lites (1992)/Kuhn et al (1994) method to the dataset

# Determine the I->QUV coefficents using just the continuum region
e = np.mean(data0[:,:,c0:c1,1]/data0[:,:,c0:c1,0])
```

```
f = np.mean(data0[:,:,c0:c1,2]/data0[:,:,c0:c1,0])
g = np.mean(data0[:,:,c0:c1,3]/data0[:,:,c0:c1,0])
# Apply the I->QUV coefficents to correct the data
data1b = np.ndarray([nZ,nY,nX,nS])
data1b[:,:,:,1] = data0[:,:,:,1] - e*data0[:,:,:,0]
data1b[:,:,:,2] = data0[:,:,:,2] - f*data0[:,:,:,0]
data1b[:,:,:,3] = data0[:,:,:,3] - g*data0[:,:,:,0]
# Do a center of mass to find the line ceneter based on the net polarized,
\hookrightarrow profile
x0 = 10-dw2
x1 = 10+dw2
xarray = np.arange(0,nX)
xarray = xarray.reshape([1,1,nX])
xcen = np.sum( np.sqrt( np.sum(data1b[:,:,x0:x1,1:]**2, axis=3) ) * xarray[:,:
 \rightarrow,x0:x1], axis=2) / \
       np.sum( np.sqrt( np.sum(data1b[:,:,x0:x1,1:]**2, axis=3) ), axis=2)
# Shift all profiles to median center position
xcen0 = int(np.median(xcen))
xshift = xcen0-xcen
data1b com = np.ndarray([nZ,nY,nX,nS])
for iZ in range(0,nZ):
    for iY in range(0,nY):
        data1b_com[iZ,iY,:,0]=shift(data1b[iZ,iY,:,0], xshift[iZ,iY], order=3,_

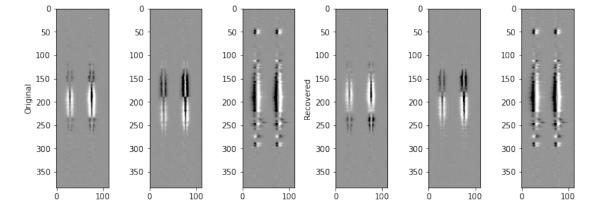
→mode='nearest')
        data1b_com[iZ,iY,:,1]=shift(data1b[iZ,iY,:,1], xshift[iZ,iY], order=3,__
 data1b com[iZ,iY,:,2]=shift(data1b[iZ,iY,:,2], xshift[iZ,iY], order=3,,,

→mode='nearest')
        data1b_com[iZ,iY,:,3]=shift(data1b[iZ,iY,:,3], xshift[iZ,iY], order=3,_
 →mode='nearest')
# Determine the Q,U->V parameters a,b through linear regression just from
→umbral profiles
x0 = 10-dw1
x1 = 10+dw1
qtmp = np.sum(data1b_com[uzidx,uyidx,x0:x1,1], axis=1)
utmp = np.sum(data1b_com[uzidx,uyidx,x0:x1,2], axis=1)
```

```
# Make a pandas dataframe to do the linear regression
     df=pd.DataFrame({'SQ':qtmp, 'SU':utmp, 'SV':vtmp}, columns=['SQ', 'SU', 'SV'])
     regr=linear_model.LinearRegression()
     regr.fit(df[['SQ','SU']], df['SV'])
     a = regr.coef [0]
     b = regr.coef_[1]
     # Make a corrected Stokes V for the determination of V->Q,U
     x0 = 10-dw2
     x1 = 10+dw2
     # Use all profiles with strong polarization signal
     qtmp = data1b_com[pzidx,pyidx,x0:x1,1]
     utmp = data1b_com[pzidx,pyidx,x0:x1,2]
     vtmp = data1b_com[pzidx,pyidx,x0:x1,3] - a*qtmp - b*utmp
     # Determine the V->QU parameters
     c = np.nanmedian( np.nanmean(qtmp*vtmp, axis=1) / np.nanmean(vtmp**2, axis=1) )
     d = np.nanmedian( np.nanmean(utmp*vtmp, axis=1) / np.nanmean(vtmp**2, axis=1) )
     # Reconstruct Sanchez Almeida & Lites (1992) I->QUV cross talk matrix
     iMM1b = np.array([[1., 0., 0., 0.],
                        [-e, 1., 0., 0.],
                        [-f, 0., 1., 0.],
                        [-g, 0., 0., 1.]], dtype='double')
     MM1b = np.linalg.inv(iMM1b)
      # Reconstruct the Kuhn et al (1994) QU<->V cross talk matrix
     0., 0.],
                                     c*b, -c],
                        [0., 1.+a*c,
                        [0., a*d, 1.+b*d, -d],
                                       -b, 1.]], dtype='double')
                        [0.,
                                 -a,
     MM2b = np.linalg.inv(iMM2b)
[11]: # Apply the final sign correction to match the original data
     iMM3b = np.array( [[1., 0., 0., 0.],
                        [0., 1., 0., 0.],
                        [0., 0., 1., 0.],
                        [0., 0., 0., 1.]], dtype='double')
     MM3b = np.linalg.inv(iMM3b)
     MMb = MM1b@MM2b@MM3b
```

vtmp = np.sum(data1b_com[uzidx,uyidx,x0:x1,3], axis=1)

```
iMMb = iMM3b@iMM2b@iMM1b
#make reconstructed data
data3b = np.einsum('ij,abcj->abci', iMMb, data0)
# Show a quick plot to verify
fig,axs = plt.subplots(ncols=6)
fig.set_size_inches(10,5)
iZ = 225 # region from the middle of the sunspot
axs[0].imshow(data[iZ,:,:,1], vmin=-0.1, vmax=0.1, cmap='Greys_r')
axs[1].imshow(data[iZ,:,:,2], vmin=-0.1, vmax=0.1, cmap='Greys_r')
axs[2].imshow(data[iZ,:,:,3], vmin=-0.1, vmax=0.1, cmap='Greys r')
axs[0].set_ylabel('Original')
axs[3].imshow(data3b[iZ,:,:,1], vmin=-0.1, vmax=0.1, cmap='Greys_r')
axs[4].imshow(data3b[iZ,:,:,2], vmin=-0.1, vmax=0.1, cmap='Greys_r')
axs[5].imshow(data3b[iZ,:,:,3], vmin=-0.1, vmax=0.1, cmap='Greys_r')
axs[3].set_ylabel('Recovered')
plt.tight_layout()
plt.show()
```



1 Look at Results

```
[12]: #Try out the inequalities from Kostinski & Givens (1993) that test the physical
→realism of a Muller matrix
#define functions for each of the quantities
def KG1(MM):
value = np.trace(MM.transpose() @ MM)
```

```
return(value)
def KG2(MM):
    mag_m1 = np.sqrt(MM[0,1]**2 + MM[0,2]**2 + MM[0,3]**2)
    mag_m2 = np.sqrt(MM[1,0]**2 + MM[2,0]**2 + MM[3,0]**2)
    value = 4.*MM[0,0]**2 + 2.*(mag_m1**2-mag_m2**2)
    return(value)
def KG3(MM):
    m1hat = MM[0,1:] / np.sqrt(np.sum(MM[0,1:]**2))
    a = MM[1:,1:]
    value = np.sum((MM[1:,0] - a@m1hat)**2)
    return(value)
def KG4(MM):
    mag_m1 = np.sqrt(MM[0,1]**2 + MM[0,2]**2 + MM[0,3]**2)
    value = (MM[0,0] - mag_m1)**2
    return(value)
print('Input Muller Matrix')
print( np.round(MMsys, decimals=3))
print(KG1(MMsys), '< or =', KG2(MMsys))</pre>
print(KG3(MMsys), '< or =', KG4(MMsys))</pre>
print('Recovered Muller Matrix, This Work')
print( np.round(MMa, decimals=3))
print(KG1(MMa), '< or =', KG2(MMa))</pre>
print(KG3(MMa), '< or =', KG4(MMa))</pre>
print('Residual Muller Matrix, This Work')
MMa_residual = np.linalg.inv(MMa) @ MMsys
print(np.round( MMa_residual, decimals=3))
tmp=np.array([np.arcsin(MMa_residual[1,2]), np.arcsin(-MMa_residual[2,1]),
              np.arccos(MMa_residual[1,1]), np.arccos(MMa_residual[2,2])])
phi=np.mean(tmp)
Sp=np.sin(phi)
Cp=np.cos(phi)
MM4a = np.array([
    [1., 0., 0., 0.],
    [0., Cp, Sp, 0.],
    [0., -Sp, Cp, 0.],
    [0., 0., 0., 1.]
], dtype='double')
iMM4a = np.linalg.inv(MM4a)
print('final geometric rotation angle', phi*180/np.pi)
print(np.round(MMa@MM4a, decimals=3))
print(np.round(np.linalg.inv(MMa@MM4a) @ MMsys, decimals=3))
```

```
print('Recovered Muller Matrix, Sanchez Almeida/Kuhn')
print( np.round(MMb, decimals=3))
print(KG1(MMb), '< or =', KG2(MMb))</pre>
print(KG3(MMb), '< or =', KG4(MMb))</pre>
print('Residual Muller Matrix, Sanchez Almeida/Kuhn')
MMb_residual = np.linalg.inv(MMb) @ MMsys
print(np.round( MMb_residual, decimals=3))
tmp=np.array([np.arcsin(MMb_residual[1,2]), -np.arcsin(MMb_residual[2,1]),
               -np.arccos(MMb_residual[1,1]), -np.arccos(MMb_residual[2,2])])
print(tmp)
phi=np.mean(tmp)
Sp=np.sin(phi)
Cp=np.cos(phi)
MM4b = np.array([
    [1., 0., 0., 0.],
     [0., Cp, Sp, 0.],
     [0., -Sp, Cp, 0.],
     [0., 0., 0., 1.]
], dtype='double')
iMM4b = np.linalg.inv(MM4b)
print('final geometric rotation angle', phi*180/np.pi)
print(np.round(MMb@MM4b, decimals=3))
print(np.round(np.linalg.inv(MMb@MM4b) @ MMsys, decimals=3))
Input Muller Matrix
[[ 1.
          0.007 0.053 -0.02 ]
 [-0.033 0.728 -0.592 0.344]
 [ 0.047  0.674  0.704 -0.22 ]
 [ 0.002 -0.112  0.392  0.911]]
4.0 < or = 4.0
0.8893218839441065 < or = 0.8893218839441067
Recovered Muller Matrix, This Work
[[ 1.
          0.022 0.049 -0.019]
[-0.033 0.528 -0.776 0.342]
 [ 0.047  0.848  0.485 -0.213]
 Γ 0.002 0.
                0.402 0.914]]
4.0 < or = 4.0
0.8893731264651672 < or = 0.8893731264651672
Residual Muller Matrix, This Work
[[ 1.
          0.
                 0.
                        0.
 Γ0.
          0.959 \quad 0.285 \quad -0.005
[ 0.
        -0.285 0.959 -0.006]
          0.003 0.007 1.
                            11
final geometric rotation angle 16.534963047739403
[[ 1.
          0.007 0.053 -0.019]
 [-0.033 0.727 -0.594 0.342]
 [ 0.047  0.675  0.706 -0.213]
 [ 0.002 -0.114  0.386  0.914]]
```

```
ΓΟ.
                     -0.
                            -0.003]
               1.
      ΓΟ.
               0.
                            -0.007]
                      1.
      Γ0.
               0.003 0.007 1.
                                  11
     Recovered Muller Matrix, Sanchez Almeida/Kuhn
     ΓΓ 1.
               0.
                      0.
                             0.
      Γ-0.033 1.
                      0.
                             0.314]
                      1.
                            -0.1927
      「 0.047 -0.
      [ 0.001 -0.377 0.248 0.834]]
     4.0379315648780425 < or = 3.9934522872877856
     nan < or = 1.0
     Residual Muller Matrix, Sanchez Almeida/Kuhn
     [[ 1.000e+00 7.000e-03 5.300e-02 -2.000e-02]
      [-0.000e+00 7.290e-01 -5.880e-01 -1.000e-03]
      [-0.000e+00 6.730e-01 7.010e-01 -9.000e-03]
      [ 1.000e-03 -4.000e-03 -4.000e-03 1.095e+00]]
     [-0.62906751 -0.73830084 -0.75336239 -0.79405267]
     final geometric rotation angle -41.75119682230743
     [[ 1.
               0.
                      0.
                             0.
                                  1
      [-0.033 0.746 -0.666 0.314]
      [ 0.047  0.666  0.746 -0.192]
      [ 0.001 -0.117  0.436  0.834]]
     [[ 1.000e+00     7.000e-03     5.300e-02     -2.000e-02]
      [-0.000e+00 9.920e-01 2.800e-02 -7.000e-03]
      [-0.000e+00 1.600e-02 9.150e-01 -6.000e-03]
      [ 1.000e-03 -4.000e-03 -4.000e-03 1.095e+00]]
     /var/folders/tk/4ztr44zd507_0fl_zvcjk8k00000gn/T/ipykernel_13349/1881087434.py:1
     4: RuntimeWarning: invalid value encountered in true_divide
       m1hat = MM[0,1:] / np.sqrt( np.sum(MM[0,1:]**2) )
[13]: # Compare the total unsigned polarization signal between recovered
      # and original Stokes vectors
      ptot_orig=np.sum( np.sqrt(np.sum(data[:,:,x0:x1,1:]**2, axis=3)) )
      ptot_new=np.sum( np.sqrt(np.sum(data3a[:,:,x0:x1,1:]**2, axis=3)) )
      ptot_kuhn=np.sum( np.sqrt(np.sum(data3b[:,:,x0:x1,1:]**2, axis=3)) )
      print(ptot_new/ptot_orig)
      print(ptot_kuhn/ptot_orig)
     0.9999996835882994
     1.0436984782758494
[14]: # Make plot showing the example spectra
      %matplotlib inline
      fig,axs=plt.subplots(ncols=4, nrows=4, sharex=True, sharey=True)
      fig.set_size_inches(10,13)
```

0. 1

ΓΓ 1.

0.

0.

```
iZ=225 #180
iY=175
y0 = 120
y1 = 270
axs[0,0].set_ylabel('Spatial Y Pixel')
axs[3,2].set_xlabel('Spectral Pixel')
axs[0,0].set title('Stokes I')
axs[0,1].set_title('Stokes Q')
axs[0,2].set title('Stokes U')
axs[0,3].set_title('Stokes V')
axs[0,0].set_ylim(y0,y1)
#axs[0,1].annotate('Original', [10,y1-20], fontsize=16, zorder=100)
iplot=axs[0,0].imshow(data[iZ,:,:,0], vmin=0, vmax=1.5, origin='lower',__

cmap='Greys_r')

pplot=axs[0,1].imshow(data[iZ,:,:,1], vmin=-0.15, vmax=0.15, origin='lower',
  ⇔cmap='Greys_r')
axs[0,2].imshow(data[iZ,:,:,2], vmin=-0.15, vmax=0.15, origin='lower',__

¬cmap='Greys_r')
axs[0,3].imshow(data[iZ,:,:,3], vmin=-0.15, vmax=0.15, origin='lower',,
  axs[0,0].plot([0,nX-1],[iY,iY], '--r')
axs[0,1].plot([0,nX-1],[iY,iY], '--r')
axs[0,2].plot([0,nX-1],[iY,iY], '--r')
axs[0,3].plot([0,nX-1],[iY,iY], '--r')
\#axs[1,0].imshow(data0[iZ,:,:,0], vmin=0, vmax=1.5, origin='lower', lower', 
  \hookrightarrow cmap = 'Greys_r')
axs[1,0].remove()
axs[1,1].imshow(data0[iZ,:,:,1], vmin=-0.15, vmax=0.15, origin='lower',__
  axs[1,2].imshow(data0[iZ,:,:,2], vmin=-0.15, vmax=0.15, origin='lower',_
  axs[1,3].imshow(data0[iZ,:,:,3], vmin=-0.15, vmax=0.15, origin='lower',__
  \#axs[2,0].imshow(data1a[iZ,:,:,0], vmin=0, vmax=1.5, origin='lower', 
  \hookrightarrow cmap = 'Greys_r')
axs[2,0].remove()
axs[2,1].imshow(data3a[iZ,:,:,1], vmin=-0.15, vmax=0.15, origin='lower', ...
   ⇔cmap='Greys_r')
```

```
axs[2,2].imshow(data3a[iZ,:,:,2], vmin=-0.15, vmax=0.15, origin='lower',__

cmap='Greys_r')

axs[2,3].imshow(data3a[iZ,:,:,3], vmin=-0.15, vmax=0.15, origin='lower',_
 ⇔cmap='Greys r')
\#axs[3,0].imshow(data2b[iZ,:,:,0], vmin=0, vmax=1.5, origin='lower', \Box
 \hookrightarrow cmap = 'Greys_r')
axs[3,0].remove()
axs[3,1].imshow(data3b[iZ,:,:,1], vmin=-0.15, vmax=0.15, origin='lower',_
 axs[3,2].imshow(data3b[iZ,:,:,2], vmin=-0.15, vmax=0.15, origin='lower',__

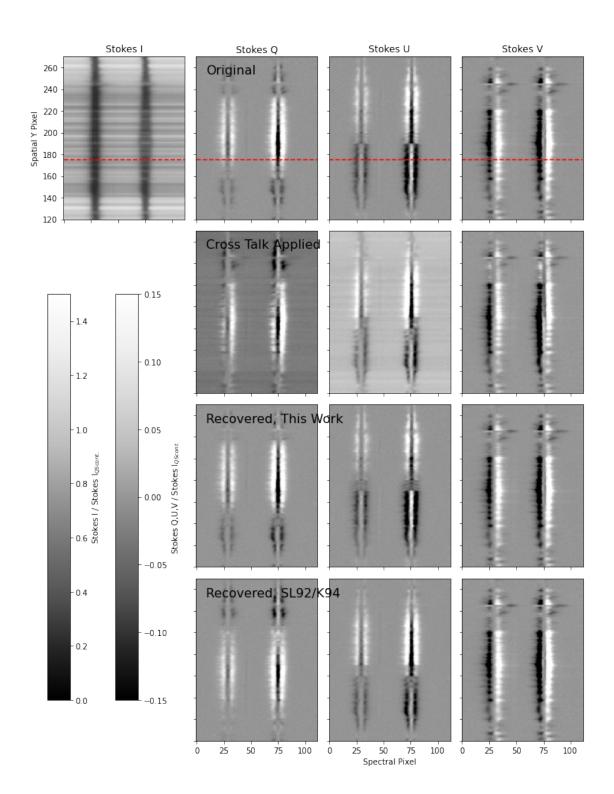
cmap='Greys_r')

axs[3,3].imshow(data3b[iZ,:,:,3], vmin=-0.15, vmax=0.15, origin='lower',

cmap='Greys_r')

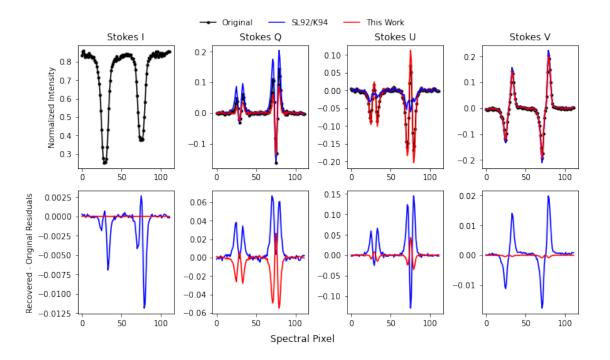
icbax = plt.axes([0.04, 0.1, 0.04, 0.55])
icb=fig.colorbar(iplot, cax=icbax, label='Stokes I / Stokes I$ {QS cont.}$')
pcbax = plt.axes([0.16, 0.1, 0.04, 0.55])
pcb=fig.colorbar(pplot, cax=pcbax, label='Stokes Q,U,V / Stokes I$_{QS cont.}$')
fig.text(0.32, 0.948, 'Original', fontsize=16)
fig.text(0.32, 0.712, 'Cross Talk Applied', fontsize=16)
fig.text(0.32, 0.476, 'Recovered, This Work', fontsize=16)
fig.text(0.32, 0.240, 'Recovered, SL92/K94', fontsize=16)
plt.tight_layout()
plt.show()
```

/var/folders/tk/4ztr44zd507_0fl_zvcjk8k00000gn/T/ipykernel_13349/2422565830.py:6
1: UserWarning: This figure includes Axes that are not compatible with
tight_layout, so results might be incorrect.
 plt.tight_layout()



[15]: # make example profile plot iZ=225 #180 iY=175

```
%matplotlib inline
fig,axs=plt.subplots(ncols=4, nrows=2)
fig.set_size_inches(10,6)
fig.supxlabel('Spectral Pixel', x=0.5, y=0.03, fontsize=12)
axs[0,0].plot(data[iZ,iY,:,0], '.-k')
axs[0,0].set title('Stokes I')
axs[0,0].set_ylabel('Normalized Intensity')
axs[0,1].plot(data[iZ,iY,:,1], '.-k', label='Original')
axs[0,1].plot(data3b[iZ,iY,:,1], 'b', label='SL92/K94')
axs[0,1].plot(data3a[iZ,iY,:,1], 'r', label='This Work')
axs[0,1].set_title('Stokes Q')
axs[0,2].plot(data[iZ,iY,:,2], '.-k')
axs[0,2].plot(data3a[iZ,iY,:,2], 'r')
axs[0,2].plot(data3b[iZ,iY,:,2], 'b')
axs[0,2].set_title('Stokes U')
axs[0,3].plot(data[iZ,iY,:,3], '.-k')
axs[0,3].plot(data3b[iZ,iY,:,3], 'b')
axs[0,3].plot(data3a[iZ,iY,:,3], 'r')
axs[0,3].set_title('Stokes V')
#axs[1,0].remove()
axs[1,0].plot(data3b[iZ,iY,:,0]-data[iZ,iY,:,0], 'b')
axs[1,0].plot(data3a[iZ,iY,:,0]-data[iZ,iY,:,0], 'r')
axs[1,0].set_ylabel('Recovered - Original Residuals')
axs[1,1].plot(data3b[iZ,iY,:,1]-data[iZ,iY,:,1], 'b')
axs[1,1].plot(data3a[iZ,iY,:,1]-data[iZ,iY,:,1], 'r')
axs[1,2].plot(data3b[iZ,iY,:,2]-data[iZ,iY,:,2], 'b')
axs[1,2].plot(data3a[iZ,iY,:,2]-data[iZ,iY,:,2], 'r')
axs[1,3].plot(data3b[iZ,iY,:,3]-data[iZ,iY,:,3], 'b')
axs[1,3].plot(data3a[iZ,iY,:,3]-data[iZ,iY,:,3], 'r')
legend = fig.legend(loc='upper center', frameon=False, ncol=3)#,__
 \rightarrow bbox_to_anchor=(0.125, 0.3))
plt.tight_layout(rect=[0.0,0.0,1.0,0.97])
plt.show()
```



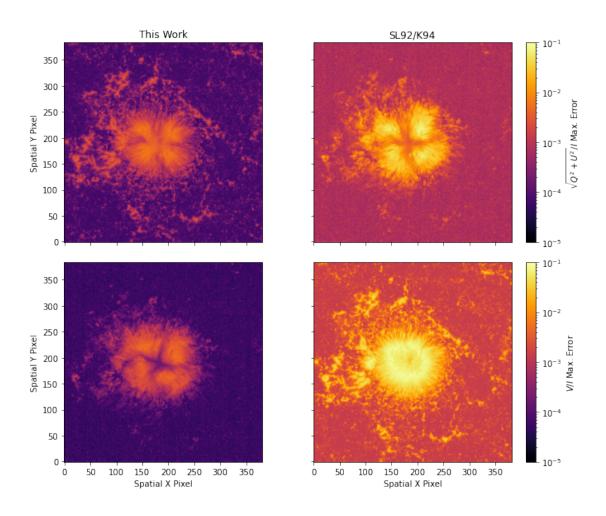
```
[16]: # Show a map of the max error in the net Stokes V polarization
                                   errQUa = np.max( np.abs(np.sqrt(data3a[:,:,x0:x1,1]**2 + data3a[:,:,x0:x1,1]**2 + data3a[:,:,x0:x1,x1]**2 + data3a[:,:,x0:x1]**2 + data3a[:,:,x0:x1]**2 + data3a[:,:,x0:x1]**2
                                          \rightarrow x1,2]**2) -
                                                                                                                                                                              np.sqrt(data[:,:,x0:x1,1]**2 + data[:,:,x0:x1,2]**2)) /__
                                          →data[:,:,x0:x1,0], axis=2)
                                   errQUa = errQUa.transpose()
                                   errVa = np.max( (np.abs(data3a[:,:,x0:x1,3]) - np.abs(data[:,:,x0:x1,3])) /__

data[:,:,x0:x1,0], axis=2)
                                   errVa = errVa.transpose()
                                   errQUb = np.max(np.abs(np.sqrt(data3b[:,:,x0:x1,1]**2 + data3b[:,:,x0:x1,1]**2 + data3b[:,:,x0:x1,x]**2 + data3b[:,:,x0:x1,x1]**2 + data3b[:,:,x0:x1]**2 + data3b[:,:,x0:x1]**2 + data3b[:,:,x0:x1]**2
                                          4x1,2]**2) -
                                                                                                                                                                              np.sqrt(data[:,:,x0:x1,1]**2 + data[:,:,x0:x1,2]**2)) /__

data[:,:,x0:x1,0], axis=2)
                                   errQUb = errQUb.transpose()
                                   errVb = np.max( (np.abs(data3b[:,:,x0:x1,3]) - np.abs(data[:,:,x0:x1,3])) /__

data[:,:,x0:x1,0], axis=2)
                                   errVb = errVb.transpose()
                                   %matplotlib inline
                                   fig,axs=plt.subplots(ncols=2, nrows=2, sharex=True, sharey=True)
                                   fig.set_size_inches(10,8)
```

```
cmap='inferno'
norm = colors.LogNorm(0.00001, 0.1, clip='True')
plotQUa=axs[0,0].imshow(errQUa, origin='lower', norm=norm, cmap=cmap)
axs[0,0].set_ylabel('Spatial Y Pixel')
axs[0,0].set_title('This Work')
plotQUb=axs[0,1].imshow(errQUb, origin='lower', norm=norm, cmap=cmap)
axs[0,1].set_title('SL92/K94')
cbarQUb=fig.colorbar(plotQUb, ax=axs[0,1], label='$\sqrt{Q^2+U^2}/I$ Max.__
 ⇔Error')
plotVa=axs[1,0].imshow(errVa, origin='lower', norm=norm, cmap=cmap)
axs[1,0].set_xlabel('Spatial X Pixel')
axs[1,0].set_ylabel('Spatial Y Pixel')
plotVb=axs[1,1].imshow(errVb, origin='lower', norm=norm, cmap=cmap)
axs[1,1].set_xlabel('Spatial X Pixel')
cbarVb=fig.colorbar(plotVb, ax=axs[1,1], label='$V/I$ Max. Error')
plt.tight_layout()
plt.show()
```



[]: