# The Crab Nebula seen through the CTA observatory

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- The Crab pulsar and its nebula is the most studied source of the Rotation Powered Pulsars (RPPs) and their nebulae, fast rotating neutron stars within the remnant of the supernova explosion (SNRe) that gave birth to the pulsars powering them.
- They are also the largest known population of gamma-ray sources in our galaxy
- Efficient particles accelerators
- Present variability in time scales from years to hours (flares).

We are interested in:

The Crab Nebula: Spectral shape + Morphology + Flares

 The Crab Pulsar: Synchrotron-curvature spectrum + Very-high-energy tail + Pulsed emission

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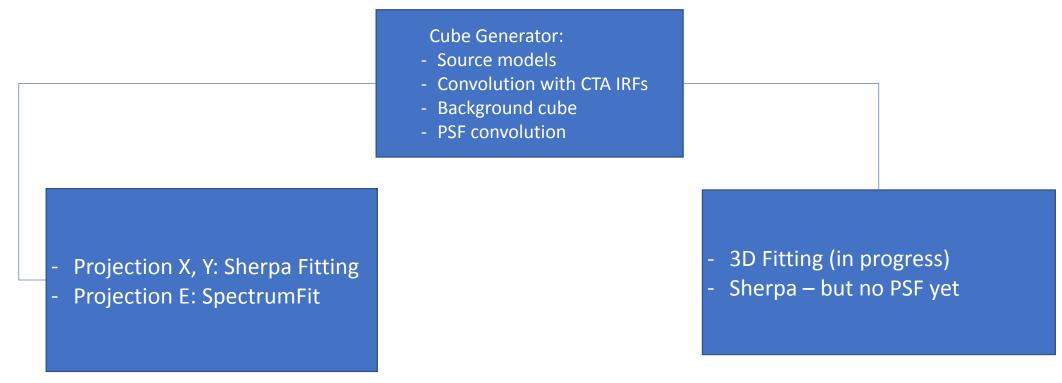
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### Methodology

- Why do we need a cube: two objects in the same position with different morphology and spectral characteristics
- Description of the simulations:



Making the cube

- 1. Nebula and pulsar models in yaml files (binning, pointing, spectrum model...):
  - Waiting for the xml implementation (generalisation to n sources)

```
def read_config(filename):
    with open(filename) as fh:
        config = yaml.load(fh)
    config['model']['prefactor'] = float(config['model']['prefactor'])
    return config

config = read_config('config.yaml')
```

- Making the cube
  - 2. Get CTA response functions:

```
def get irfs(config):
   filename = 'irf file.fits'
   offset = Angle(config['selection']['offset fov'] * u.deg)
   psf fov = EnergyDependentMultiGaussPSF.read(filename, hdu='POINT SPREAD FUNCTION')
   psf = psf fov.to energy dependent table psf(theta=offset)
   aeff = EffectiveAreaTable2D.read(filename, hdu='EFFECTIVE AREA')
   edisp fov = EnergyDispersion2D.read(filename, hdu='ENERGY DISPERSION')
   table = fits.open(filename)['BACKGROUND']
   table.columns.change name(str('BGD'), str('Bgd'))
   table header['TUNIT7'] = '1 / (MeV s sr)'
   bkg = Background3D.read(filename, hdu='BACKGROUND')
    return dict(psf=psf, aeff=aeff, edisp=edisp_fov, bkg=bkg, offset = offset)
irfs = get irfs(config)
```

- Making the cube
  - 3. Make a CombineModel3D:

```
def get model gammapy(config):
  if config['model']['template'] == 'Shell2D':
        spatial_model = Shell2D(
           amplitude=1,
           x_0=config['model']['ra'],
           v 0=config['model']['dec'],
            r in=config['model']['rin'],
           width=config['model']['width'],
           # Note: for now we need spatial models that are normalised
           # to integrate to 1 or results will be incorrect!!!
           normed=True,
  if config['model']['spectrum'] == 'pl':
        spectral_model = PowerLaw(
            amplitude=config['model']['prefactor'] * u.Unit('cm-2 s-1 TeV-1'),
           index=config['model']['index'],
           reference=config['model']['pivot_energy'] * u.Unit('TeV'),
  return CombinedModel3D(
        spatial_model=spatial_model,
        spectral model=spectral model,
  model = get_model_gammapy(config)
```

- Making the cube
  - 4. Make reference cubes:

```
def make_ref_cube(config):
    WCS SPEC = {
        'nxpix': config['binning']['nxpix'],
        'nypix': config['binning']['nypix'],
        'binsz': config['binning']['binsz'],
        'xref': config['pointing']['ra'],
        'yref': config['pointing']['dec'],
        'proj': config['binning']['proj'],
        'coordsys': config['binning']['coordsys'],
   # define reconstructed energy binning
    ENERGY_SPEC = {
        'mode': 'edges',
        'enumbins': config['binning']['enumbins'],
        'emin': config['selection']['emin'],
        'emax': config['selection']['emax'],
        'eunit': 'TeV'.
    CUBE SPEC = \{\}
    CUBE_SPEC.update(WCS_SPEC)
    CUBE SPEC.update(ENERGY SPEC)
    cube = SkyCube.empty(**CUBE SPEC)
    return cube
ref_cube = make_ref_cube(config)
```

- Making the cube
  - 5. Evaluate the model in the reference cubes and sum the fluxes:

```
flux cube = model.evaluate cube(ref cube)
flux cube2 = model2.evaluate cube(ref cube2)
def compute sum_cube(flux_cube, flux_cube2, config):
    ebin = flux cube.energies(mode="edges")
    ebounds = EnergyBounds(ebin)
    nflux cube sum = make ref cube(config)
    for idx in range(len(ebounds) - 1):
        npred1 = flux cube.sky image idx(idx)
        npred2 = flux cube2.sky image idx(idx)
        nflux_sum = u.Quantity(npred1.data.value + npred2.data.value,'1 / (cm2 s sr TeV)')
        nflux cube sum.data[idx] = nflux_sum.value
    return nflux_cube_sum
flux sum=compute sum cube(flux cube, flux cube2, config)
```

Making the cube

#### 6. Make predicted counts cube

```
def compute_npred_cube(flux_cube, exposure_cube, ebounds, config, irfs, integral_resolution=10):
   # Make an empty cube with the requested energy binning
    sky_geom = exposure_cube.sky_image_ref
    energies = EnergyBounds(ebounds)
   npred_cube = SkyCube.empty_like(sky_geom, energies=energies, unit='', fill=np.nan)
   # Process and fill one energy bin at a time
   for idx in range(len(ebounds) - 1):
       emin, emax = ebounds[idx: idx + 2]
       ecenter = np.sqrt(emin * emax)
       flux = flux cube.sky image integral(emin, emax, interpolation='linear', nbins=integral resolution)
       exposure = exposure cube.sky image(ecenter, interpolation='linear')
       solid angle = exposure.solid angle()
       flux.data.value[np.isnan(flux.data.value)]=0
       exposure.data.value[np.isnan(exposure.data.value)]=0
       npred = flux.data.value * u.Unit('1 / (cm2 s sr)') * exposure.data * solid angle
       npred cube data[idx] = npred to('')
   # Apply EnergyDispersion
   edisp = irfs['edisp']
    offset = irfs['offset']
   edisp_idx = edisp.to_energy_dispersion(offset=offset, e_reco = ebounds, e_true = ebounds)
   for pos x in range(npred cube.data.shape[1]):
        for pos_y in range(npred_cube.data.shape[2]):
           npred_pos = npred_cube.data[0:len(ebounds) - 1,pos_x,pos_y]
           if npred pos.sum() != 0:
              for idx in range(len(ebounds) - 1):
                 npred_cube.data[idx][pos_x][pos_y] = np.dot(npred_pos, edisp_idx.data.data[idx])
    return npred_cube
```

- Making the cube
  - 7. Compute the on and off events and the excess

```
def compute nexcess cube(npred cube, livetime, pointing, offset max, bkg rate, config):
   ebin = npred cube.energies(mode="edges")
   ebounds = EnergyBounds(ebin)
   nexcess cube = make ref cube(config)
   non cube = make ref cube(config)
   noff cube = make ref cube(config)
   # Compute two background cubes
   nbkq1 cube = make background cube(pointing = pointing, obstime = livetime, bkg = bkg rate, ref cube=npred cube, offset max = offset max)
   nbkg2 cube = make background cube(pointing = pointing, obstime = livetime, bkg = bkg rate, ref cube=npred cube, offset max = offset max)
   # For each energy bin, I need to obtain the correct background rate (two, one for the on and one for the off)
   for idx in range(len(ebounds) - 1):
        emin, emax = ebounds[idx: idx + 2]
        ecenter = np.sqrt(emin * emax)
       npred = npred_cube.sky_image_idx(idx)
       npred.unit = u.Unit('TeV')
        solid_angle = npred.solid_angle()
       npred.data.value[np.isnan(npred.data.value)]=0.
       nbkg1 ebin = nbkg1 cube.data[idx]
       nbkg2 ebin = nbkg2 cube.data[idx]
       n on = np.random.poisson(npred.data) + np.random.poisson(abs(nbkg1 ebin))
       n_off = np.random.poisson(abs(nbkg2_ebin))
       nexcess = n on - n off
       nexcess cube.data[idx] = nexcess
       non cube.data[idx] = n on
       noff cube.data[idx] = n off
   return nexcess cube, non cube, noff cube
```

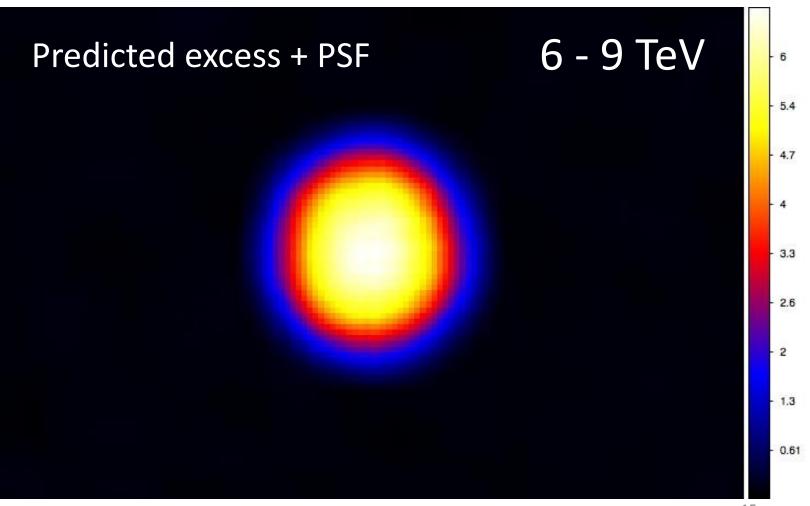
- Making the cube
  - 8. Apply PSF convolution and write the cubes

```
kernels = []
psf_table = psf_fromfits('irf_file.fits')
energ lo = psf table[0].value
sigmas = psf table[3]
energ_array = nexcess_cube.energies('edges')
s = np.argmin(np.abs(irfs['offset'].value - psf_table[2].value))
for i in range(nexcess cube.data.shape[0]):
   v = np.argmin(np.abs(energ_array[i].value - energ_lo))
    kernels.append(irfs['psf'].kernels(nexcess cube,Angle(sigmas[0][s][v] * u.deg))[i])
# Apply kernels convolution
nexcess_cube_convolved = nexcess_cube.convolve(kernels)
noff cube convolved = noff cube.convolve(kernels)
non_cube_convolved = non cube.convolve(kernels)
nexcess_cube_convolved.write('nexcess_cube_convolved.fits.gz', overwrite = True)
noff cube convolved.write('noff cube convolved.fits.gz', overwrite=True)
non_cube_convolved.write('non cube convolved.fits.qz', overwrite = True)
```

• Example of cube. Excess events cube convolved with the PSF:

#### Model:

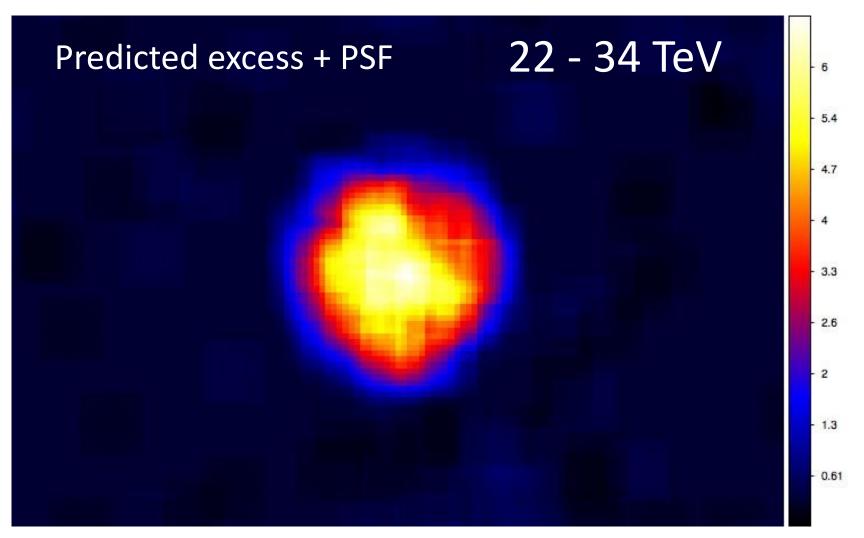
- MAGIC Log-Parabola
- 2D sphere of 0.1 deg
- Pulsar index of 2.9
- Point-like pulsar
- Time of 100 hours



• Example of cube. Excess events cube convolved with the PSF:

Same cube at higher energies:

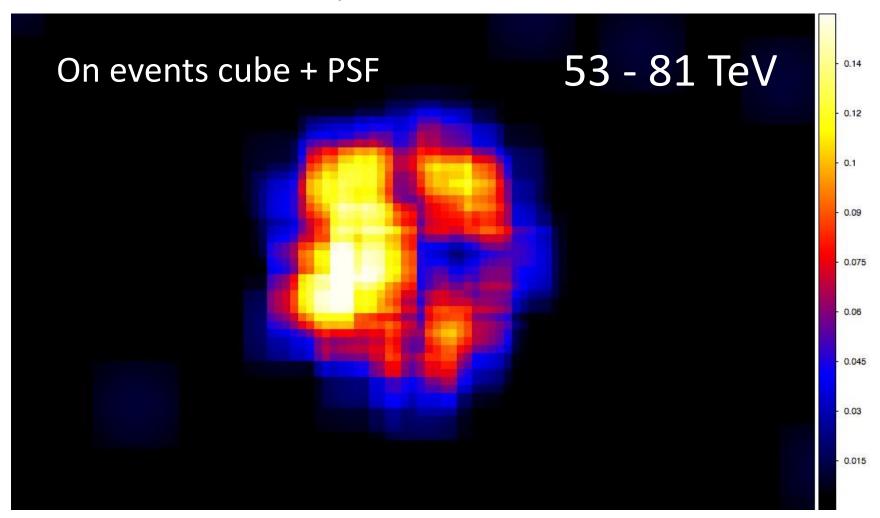
Effect of the PSF energy dependence + on events spatial fluctuations



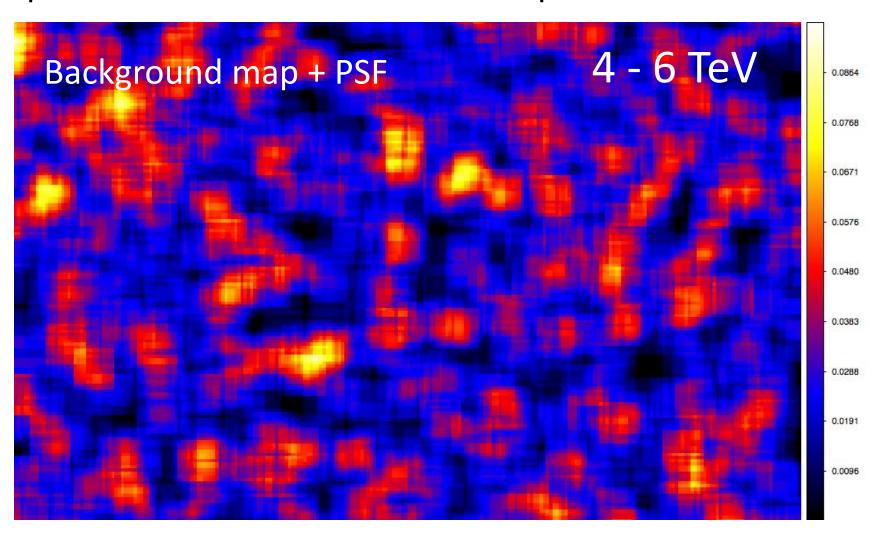
• Example of cube. On and off events maps convolved with PSF:

Same cube at higher energies:

Effect of the PSF energy dependence + on events spatial fluctuations



• Example of cube. On and off events maps convolved with PSF:



• For the moment we project the cube along the E axis to fit the spectrum, we can: Switch the pulsar off (nebula + background) or switch the nebula off (pulsar + background), so for example we take the nebula as part of the background cube in order to fit the pulsar spectrum.

```
filename_on = 'non_cube_convolved.fits.gz'
cube_on = SkyCube.read(filename_on)
circ_filename_off = 'noff_withneb_cube_convolved.fits.gz'
circ_cube_off = SkyCube.read(circ_filename_off)
```

1. Define on and off regions by energy bins.

```
filename on = 'non cube convolved.fits.gz' # non cube convolved.fits
cube on = SkyCube.read(filename on)
circ filename off = 'noff withneb cube convolved.fits.gz'
circ cube off = SkyCube.read(circ filename off)
psf_table = psf_fromfits('irf_file.fits')
psfs = psf table[3]
on sizes = np.ones(int(config['binning']['enumbins'])) * u.deg
energarr = cube on.energies('edges')
for idx in range(len(cube on energies('center'))):
    i = np.argmin(np.abs(energarr[idx].value - psf_table[0].value))
    j = np.argmin(np.abs(offset_fov - psf_table[2].value))
    on sizes.value[idx] = psfs[0][i][i]
on pos = SkyCoord(83.6333 * u.deg, 22.0144 * u.deg, frame='icrs')
off pos = SkyCoord(83.6333 * u.deg, 22.0144 * u.deg, frame='icrs')
```

#### 2. Take counts spectrum:

```
on_region = CircleSkyRegion(on_pos, on_sizes[i])
off_region = CircleSkyRegion(off_pos, off_sizes[i])
on_data['value'][i] = cube_on.spectrum(on_region)['value'][i]
off_data['value'][i] = cube_off.spectrum(off_region)['value'][i]
```

### 3. Apply some criteria to define the fitting range:

```
if limasig >= 3 and on_data['value'][i] - off_data['value'][i] >= 7:

circ_on_vector = PHACountsSpectrum(energy_lo = cube_on.energies('edges')[:-1], energy_hi= cube_on.energies('edges')[1:],
data= circ_on_data['value'].data * circ_stats * u.ct, backscal = on_sizes[0].value, meta={'EXPOSURE' : livetime.value})

circ_off_vector = PHACountsSpectrum(energy_lo = circ_cube_off.energies('edges')[:-1], energy_hi=
circ_cube_off.energies('edges')[1:], data= circ_off_data['value'].data * circ_stats * u.ct, backscal = off_sizes[0].value,
meta={'EXPOSURE' : livetime.value, 'OFFSET' : 0.3 * u.deg})
```

#### 4. Make spectrum observation and fit

```
circ_sed_table = SpectrumObservation(on_vector = circ_on_vector, off_vector = circ_off_vector, aeff = aeff,
edisp = edisp)

fit_source = SpectrumFit(obs_list = ann_sed_table, model=models_ann_fit[k],forward_folded=True,
fit_range=(0.05 * u.Unit('TeV'), cube_on.energies('edges')[int(np.sum(ann_stats))]))
fit_source.fit()
fit_source.est_errors()
results = fit_source.result
```

#### 5. Add systematical errors

- Main source of errors
- For now relaying in the current literature (MAGIC, HESS), supposing some factor of improvement (factor of 2 for example)

- For now relaying in the current literature (MAGIC, HESS):

Systematical errors (%)

Parameter \ Model	LogParabola	ExponentialPowerLaw	PowerLaw
${ m E_0}$	17	17	17
$N_0$	11	20	20
$\mathbf{Index}$		4	2
Cutoff Energy		20	
lpha	2		
$oldsymbol{eta}$	30		

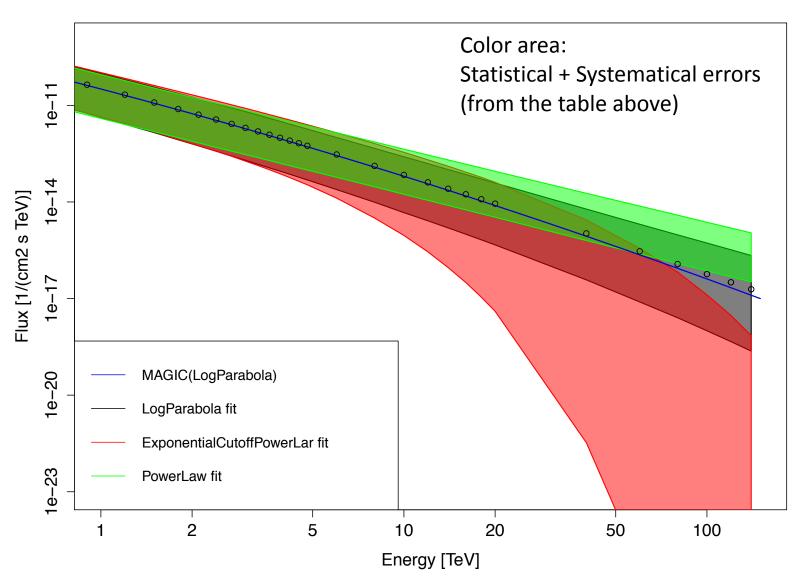
How much does the spectrum extends?

Time to detect the nebula at 5 sigma for the energy range displayed

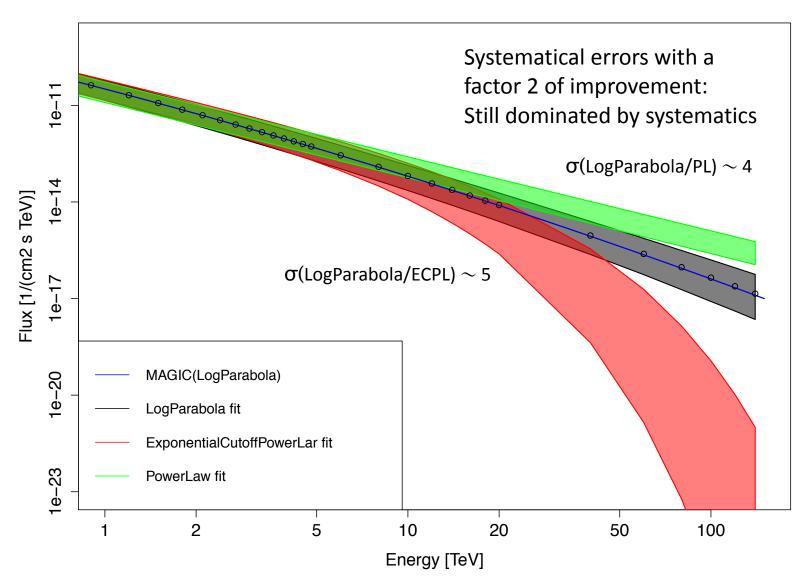
MAGIC (Log-Parabola)		HESSII (Log-Parabola)	
Energy (TeV)	$t(\sigma = 5)$ [h]	Energy (TeV)	$t(\sigma = 5)$ [h]
E < 0.1	0.37	E < 0.1	3.8
0.1 < E < 1	0.0056	0.1 < E < 1	0.007
E > 5	0.17	E > 5	0.22
E > 50	30.3	E > 50	232.8

How much does the spectrum extends? What is the spectral shape?

#### CTA North 50h



#### CTA North 50h

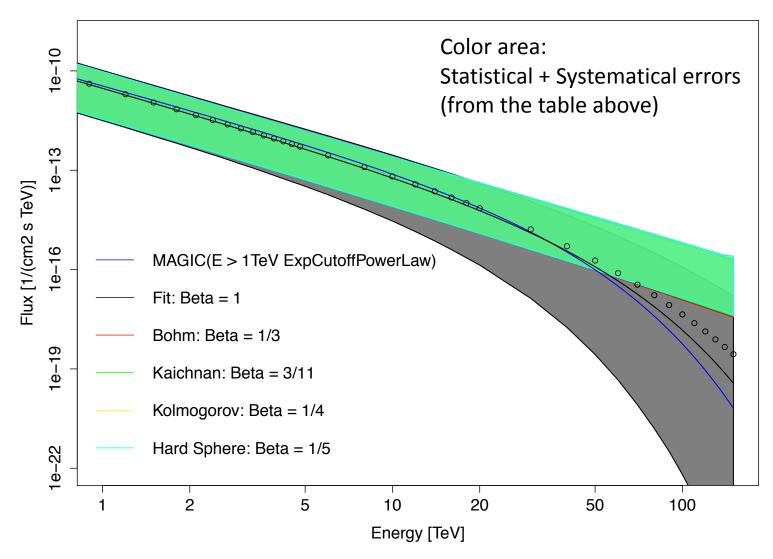


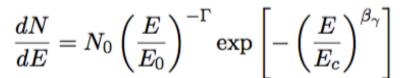
How much does the spectrum extends? What is the spectral shape?
 How is the shape of the spectral cut-off?

$$rac{dN}{dE} = N_0 \left(rac{E}{E_0}
ight)^{-\Gamma} \exp\left[-\left(rac{E}{E_c}
ight)^{eta_\gamma}
ight]$$

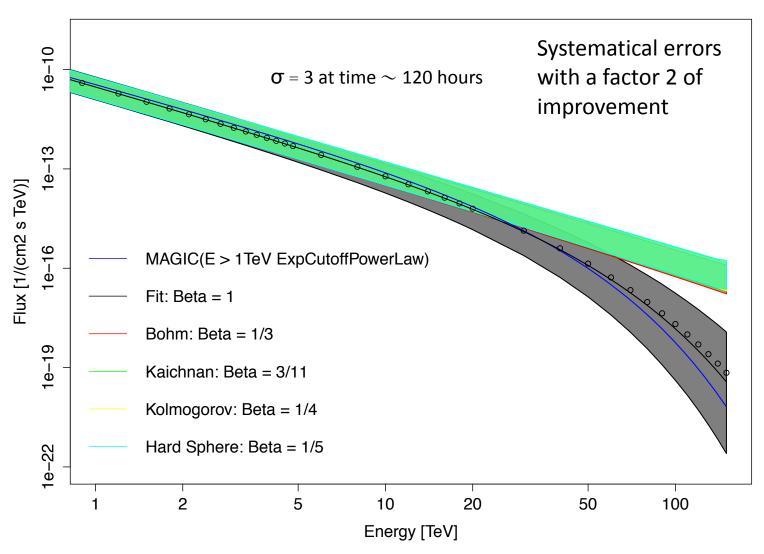
 $rac{dN}{dE} = N_0 \left(rac{E}{E_0}
ight)^{-\Gamma} \exp\left[-\left(rac{E}{E_c}
ight)^{eta_\gamma}
ight]$ 

CTA North 50h





CTA North 50h



### **Spectral Analysis: Conclusions**

- We needed at least a factor of two of improvement for the systematical errors in order to distinguish between the input and the other models.
- We can recover the initial spectral model with small statistical errors in comparison with systematics.
- We could distinguish sub-exponential cutoff models with differences in the  $\beta$  parameter greater than 0.67.

#### 1. Introduce the PSF

```
def psf fromfits(filename):
    hdulist = pyfits.open(filename)
    hdu = hdulist[2]
    energy_lo = Quantity(hdu.data['ENERG_LO'][0], 'TeV')
    energy_hi = Quantity(hdu.data['ENERG_HI'][0], 'TeV')
    theta = Angle(hdu.data['THETA LO'][0], 'deg')
    # Get sigmas
     shape = (len(theta), len(energy hi))
                                                                 For the moment we just use a
     sigmas = []
     for key in ['SIGMA 1', 'SIGMA 2', 'SIGMA 3']:
                                                                 simple gaussian with \sigma_1
        sigma = hdu.data[key].reshape(shape).copy()
        sigmas.append(sigma)
    # Get amplitudes
     norms = []
     for key in ['SCALE', 'AMPL_2', 'AMPL_3']:
        norm = hdu.data[key].reshape(shape).copy()
        norms_append(norm)
     return dict(energy_lo = energy_lo, energy_hi = energy_hi, theta = theta, sigmas = sigma, norms =
norms)
  psf_table = psf_fromfits('irf_file.fits')
```

#### 2. Fit morphology for each energy bin I

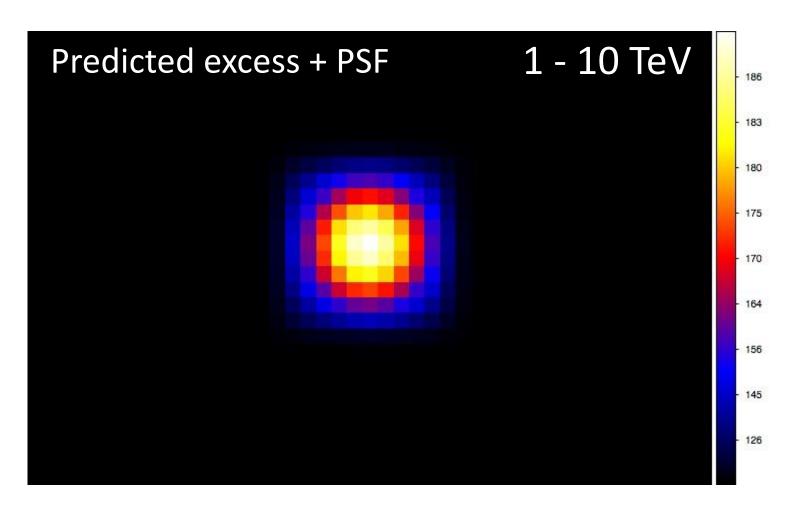
```
filename = 'nexcess_cube_convolved.fits.qz'
                                                    ## Read file to fit
cube = SkyCube.read(filename)
config = read config('config.yaml')
binsz = config['binning']['binsz']
offset fov = config['selection']['offset fov']
# Take PSF data
irffile = 'irf file.fits'
psf_table = psf_fromfits(irffile)
energarr = cube.energies('edges')
sigmas = psf table[3]
norms = psf table[4]
hdu = pyfits.open(filename)
im sizex = hdu[0] header['NAXIS1']
im sizey = hdu[0].header['NAXIS2']
cx = 0.5*im sizex
cy = 0.5*im^-sizey
# Check the significance...
# Make image cube from slice excess convolved cube
cube sum = np.zeros((cube.data.shape[1],cube.data.shape[2])) * u.ct
cube_sum = np.add(cube_sum, cube.data[idx])
image sum = SkyCube.empty like(cube)
image sum data = cube sum
image sum.write('sum image.fits.gz', overwrite=True)
# Find nearest energy and theta value
i = np.argmin(np.abs(energarr[idx].value - psf table[0].value))
j = np.argmin(np.abs(offset fov - psf table[2].value))
# Make PSF
s1 = sigmas[0][j][i]/binsz
s2 = sigmas[1][i][i]/binsz
s3 = sigmas[2][j][i]/binsz
ampl = norms[0][i][i]
ampl2 = norms[1][j][i]
ampl3 = norms[2][i][i]
```

3. Fit morphology for each energy bin II. We use sherpa

```
# Morphological fitting
load image("sum image.fits.gz")
set method("simplex")
set stat("cash")
x0 = 125
y0 = 125
rad0 = 48.0
image getregion(coord="physical")
'circle(x0,y0,rad0);'
notice2d("circle(" + str(x0) + "," + str(y0) + "," + str(rad0) + ")")
load user model(GaussianSource, "sph2d")
add_user_pars( "sph2d", ["sigma1", "sigma2", "sigma3", "alpha", "beta", "ampl", "size", "xpos", "ypos"] )
set model(sph2d + const2d.bgnd)
show model()
fit()
conf()
save model("model " + str(idx) + ".fits")
save resid("resid_" + str(idx) + ".fits")
```

• Example of cube. Excess convolved with PSF:

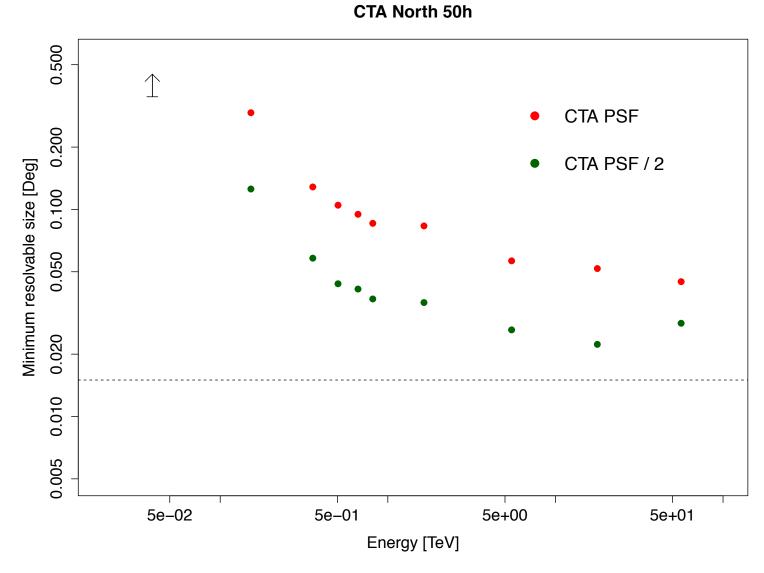
- Model:
  - MAGIC Log-Parabola
  - 2D sphere 52"
  - Pulsar index: 2.9
  - Point-like pulsar
  - Time of 30 hours
  - PSF ~ 2'
- Question: To recover the TeV source extension (Sherpa)



 We can either fix the CTA IRFs and change the size of the source until the one is resolved:

#### • Model:

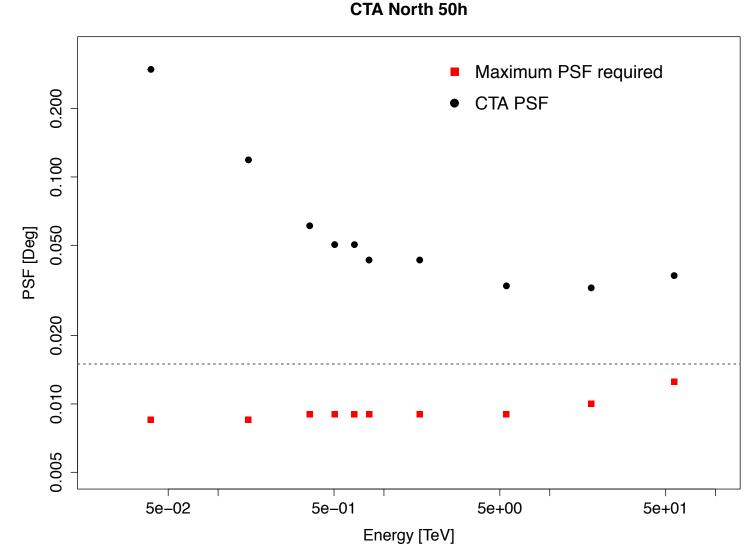
- MAGIC Log-Parabola
- 2D sphere 52' (dashed line)
- Pulsar index: 2.9
- Point-like pulsar
- Time of 50 hours



 Or fix the source size and change the PSF convolved in the cube generator until it is resolved:

#### Model:

- MAGIC Log-Parabola
- 2D sphere 52' (dashed line)
- Pulsar index: 2.9
- Point-like pulsar
- Time of 50 hours



### To conclude:

- The cube generation allow us to place simulations of two objects in the same position with different morphology and spectral characteristics plus the background.
- From the simulated cubes it is possible to make spectral and morphological analysis with gammapy and sherpa.
- We can apply this tools to look for what to expect of the Crab Nebula and Pulsar regarding to; detection significance, spectral model or morphology with the CTA observatory