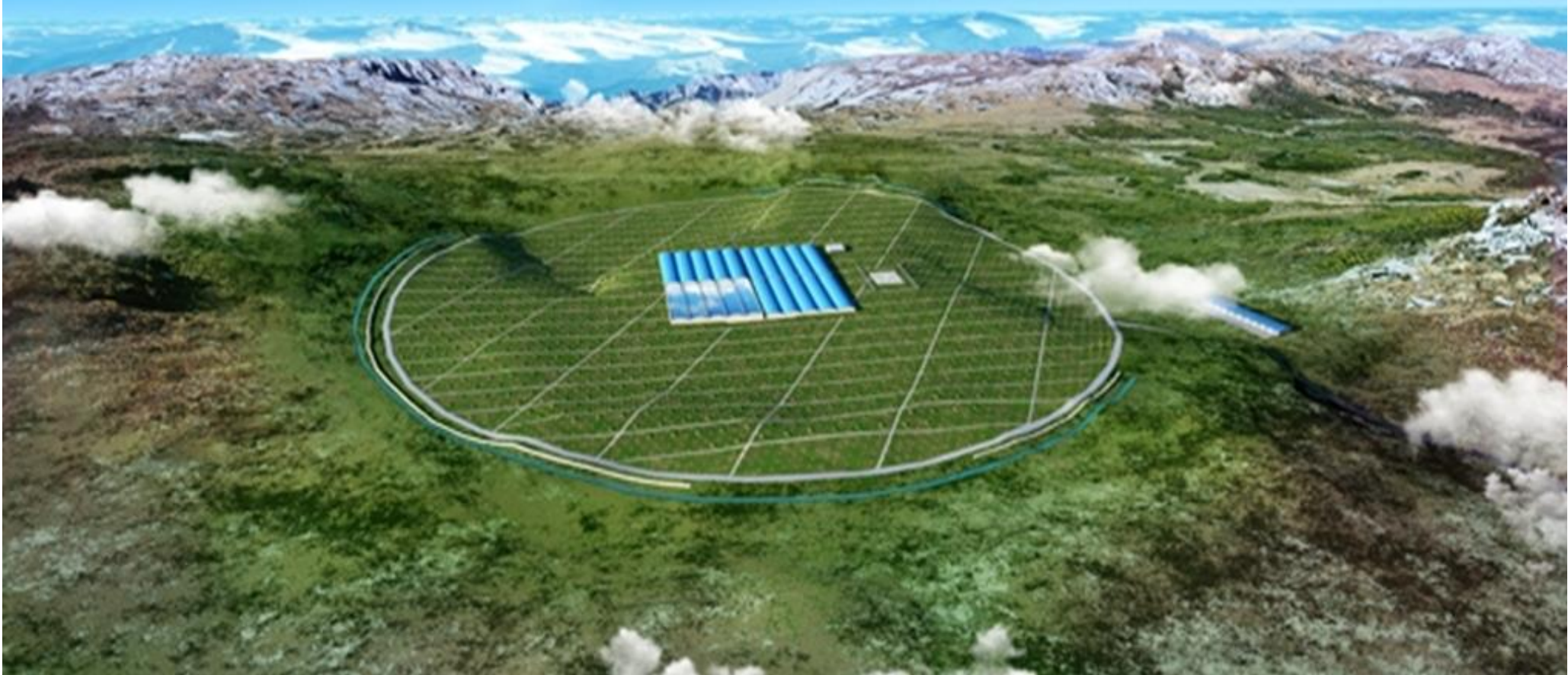


A 3D Likelihood Analysis Tool for LHAASO-KM2A data

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What's 3D likelihood

Spectrum + Spatial distribution Fit

- Space-base: Fermi-LAT, DAMPE
- Ground-base: HWAC, HESS, CTA

The source model is considered as:

$$S(E, \hat{p}, t) = \sum_i s_i(E, t) \delta(\hat{p} - \hat{p}_i) + S_G(E, \hat{p}) + S_{eg}(E, \hat{p}) + \sum_l S_l(E, \hat{p}, t),$$

Point Sources

Galactic & EG Diffuse Sources

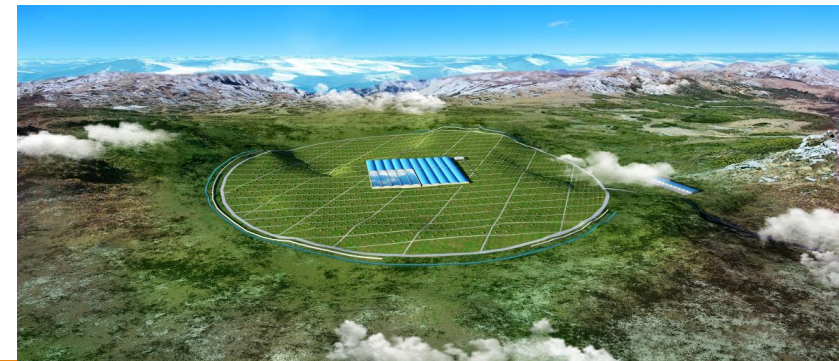
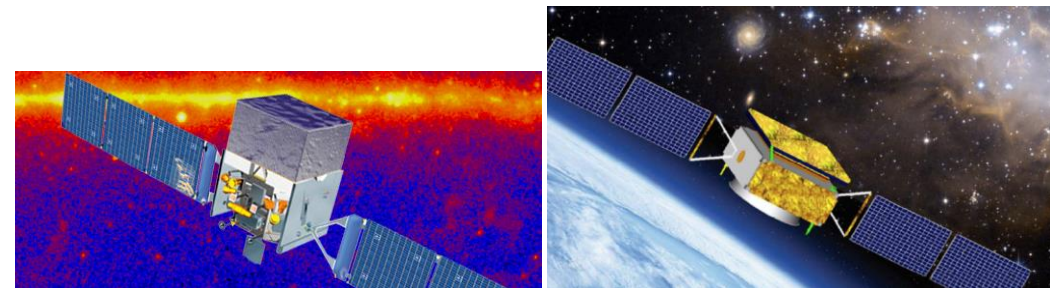
Other Sources

This model is folded with the Instrument Response Functions (IRFs) to obtain the predicted counts in the measured quantity space (E', p', t') :

$$M(E', \hat{p}', t) = \int_{SR} dE d\hat{p} R(E', \hat{p}', t; E, \hat{p}) S(E, \hat{p}, t)$$

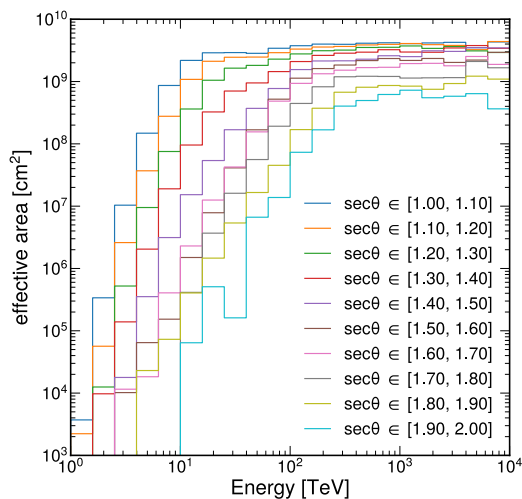
where

$$R(E', \hat{p}'; E, \hat{p}, t) = A(E, \hat{p}, \vec{L}(t)) D(E'; E, \hat{p}, \vec{L}(t)) P(\hat{p}'; E, \hat{p}, \vec{L}(t))$$

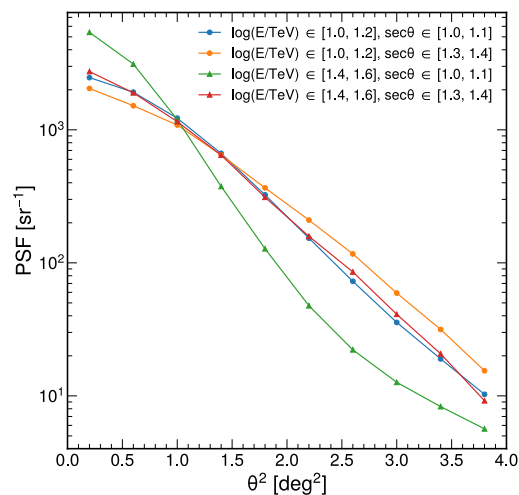


KM2A IRFs from Simulation

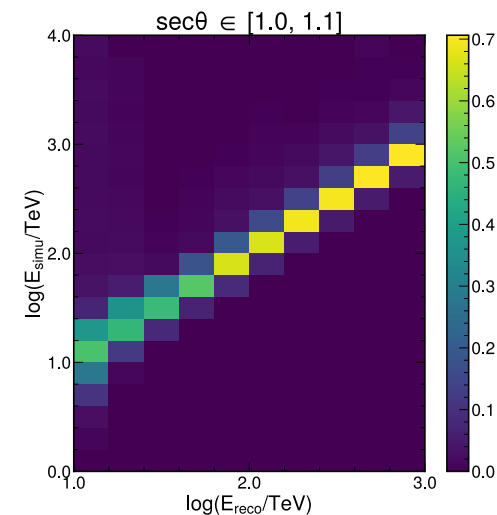
Instrument Response Functions (**IRFs**) including the effective area, point-spread function and energy dispersion represent the performance of the detections like sensitivity, angular and energy resolution.



Effective Area

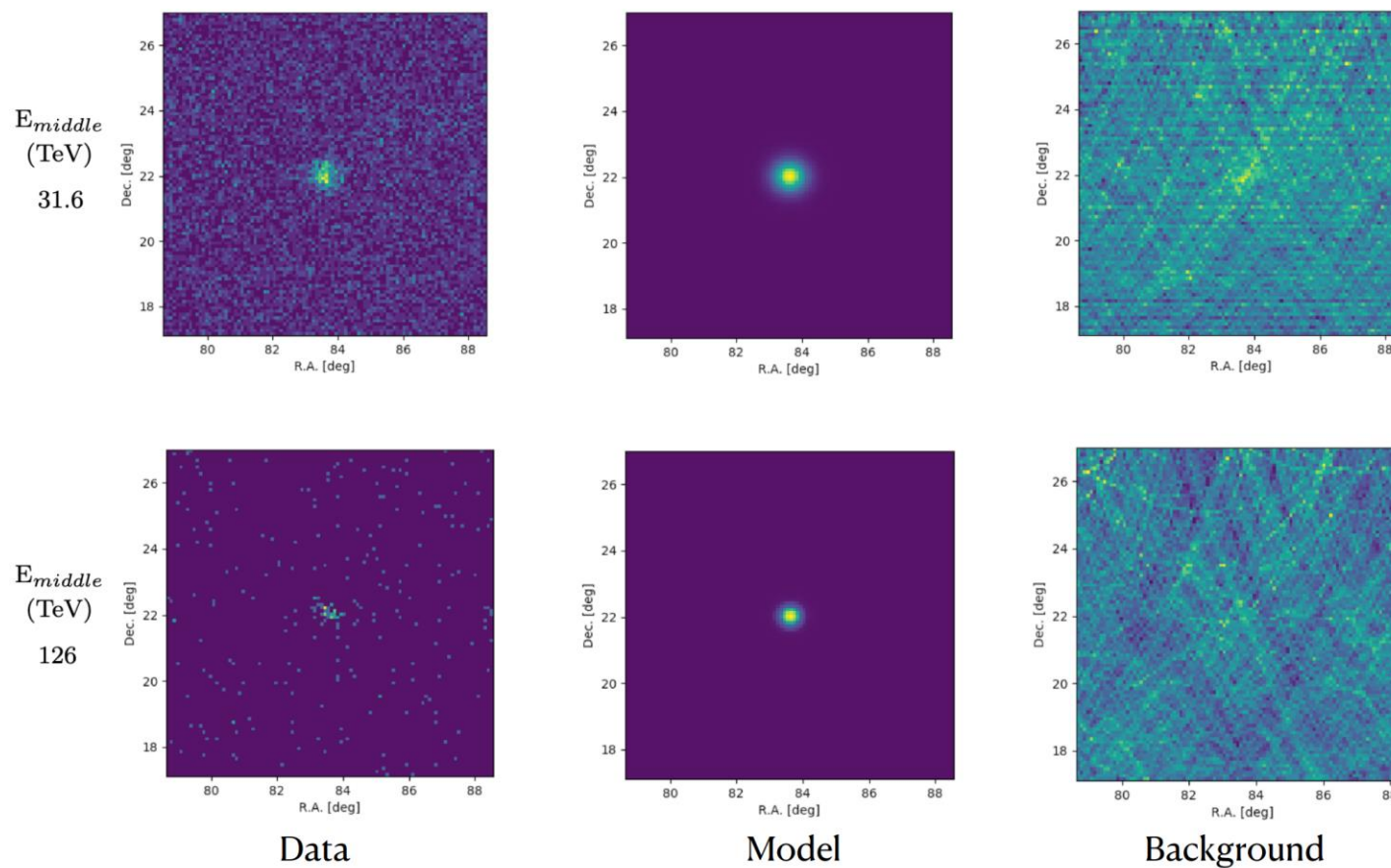


Point Spread Function

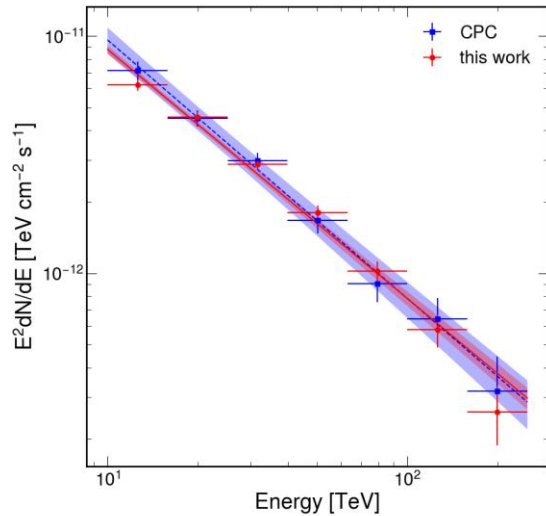


Energy Dispersion

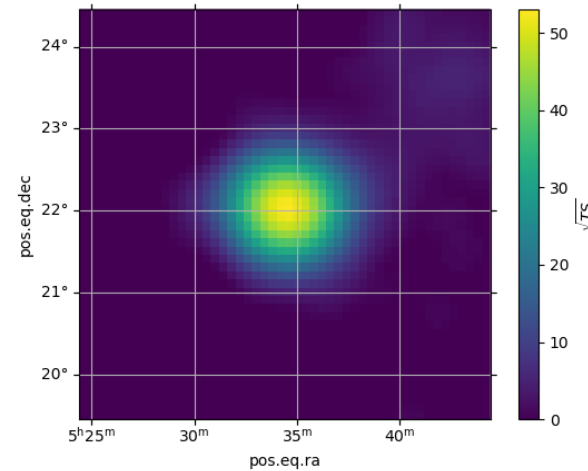
KM2A Data and Model Prediction



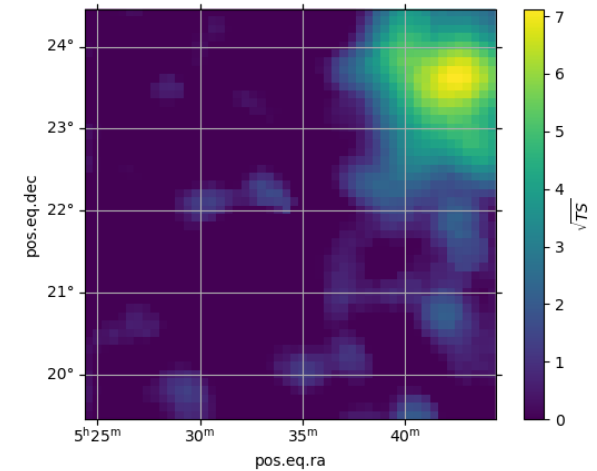
Analysis for Point Source



Spectral Energy Distribution (SED)



Test Statistic (TS) map



residual TS map

The analysis with this software could give consistent results with those using traditional method.

Analysis for Extended Source

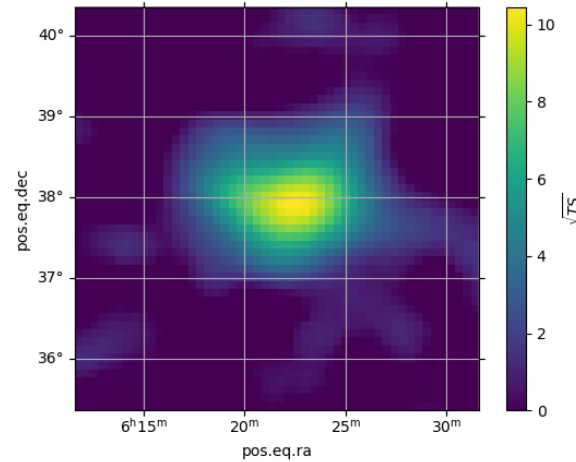
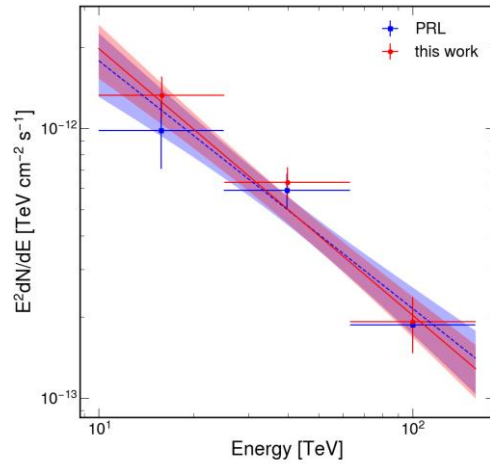


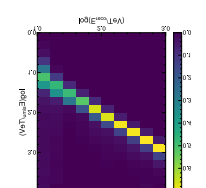
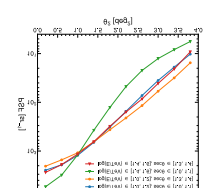
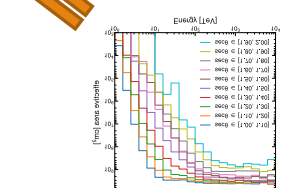
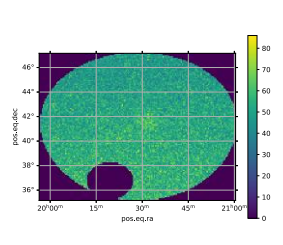
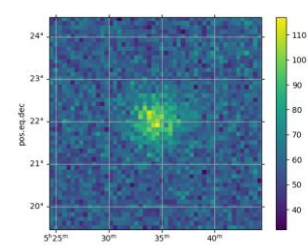
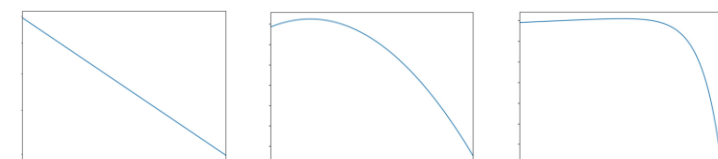
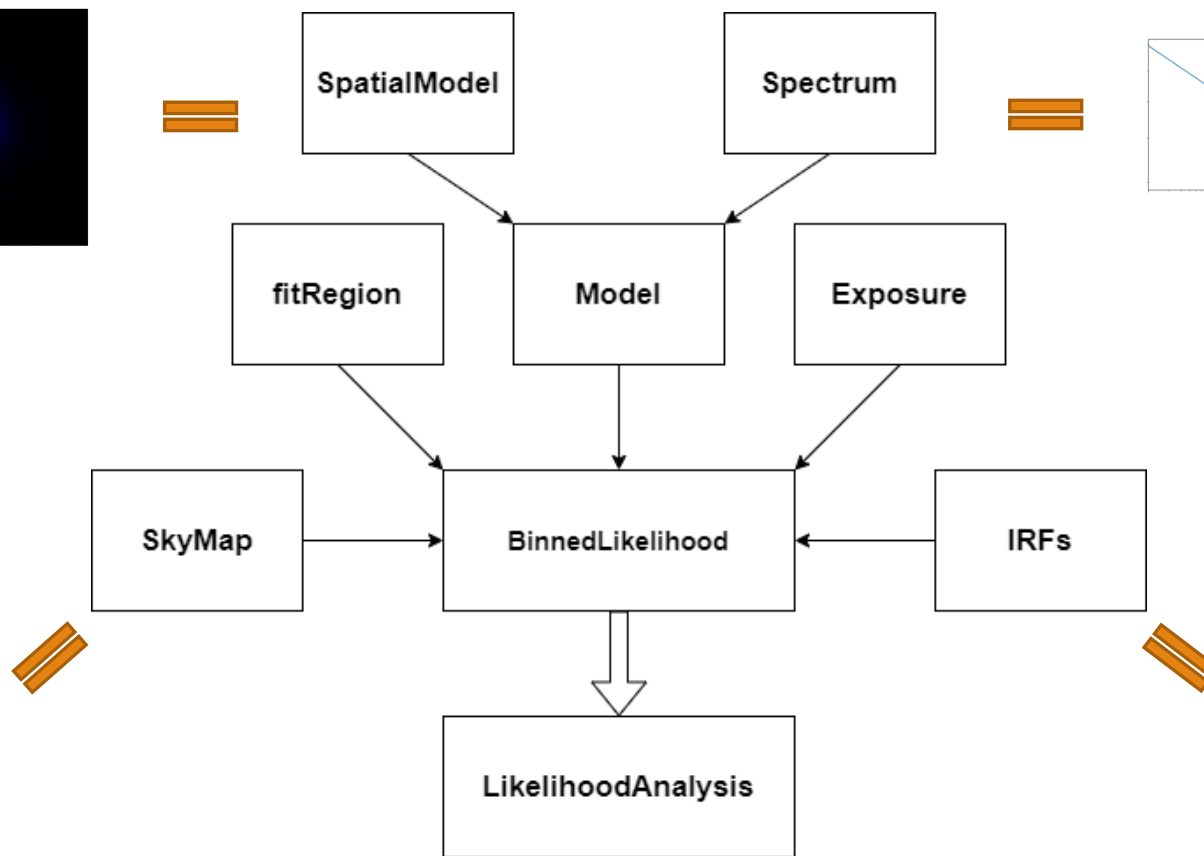
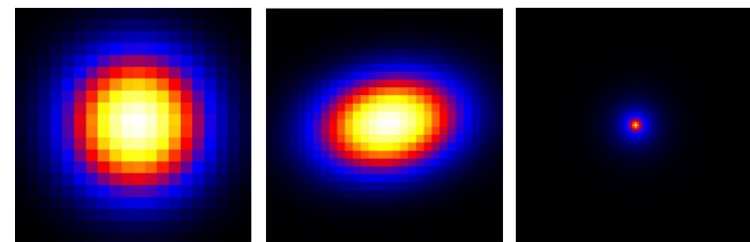
TABLE I. Results of the morphological analyses of LHAASO J0621+3755.

Template	Extension ^a (°)	RA (°)	Dec (°)	TS	N_p^b
Point source	-	95.56 ± 0.10	37.85 ± 0.07	63.0	3
2D Gaussian	0.40 ± 0.07	95.47 ± 0.11	37.92 ± 0.09	79.5	4
Uniform disk	0.70 ± 0.10	95.44 ± 0.11	37.94 ± 0.09	80.2	4
Diffusion	0.91 ± 0.20	95.48 ± 0.10	37.90 ± 0.09	78.1	4

^aRadius for the uniform disk; σ for the Gaussian model; θ_d for the diffusion model. ^b N_p is the number of parameters in the model.

Template	Extension	RA	DEC	Prefactor	Index	TS
Point source	-	95.62+/-0.09	37.92+/-0.06	1.83+/-0.28	2.89+/-0.18	95.1
2D Gaussian	0.36+/-0.06	95.52+/-0.09	37.89+/-0.07	3.0+/-0.4	2.99+/-0.15	133.8
Uniform disk	0.65+/-0.12	95.51+/-0.08	37.92+/-0.07	2.9+/-0.4	3.01+/-0.15	130.9
Diffusion	0.82+/-0.17	95.54+/-0.08	37.88+/-0.06	3.1+/-0.4	2.99+/-0.14	125.0

Implementation



For Users - Input

SpatialModel:

- Point Source
 - SkyDirFunction: RA, DEC
- Extended Source
 - RadialDisk: RA, DEC, Radius
 - RadialGaussian: RA, DEC, Sigma
 - Ring2D: RA, DEC, R_in, R_out
 - Ellipse2D: RA, DEC, A, B, Theta
 - Gaussian2D: RA, DEC, X_stddev, Y_stddev, Theta
 - PulsarHalo: RA, DEC, ThetaD
- SpatialMap
- MapCubeFunction

Spectrum:

- PowerLaw: Prefactor, Index, Scale
- BrokenPowerLaw: Prefactor, Index1, Index2, BreakValue
- LogParabola: norm, alpha, beta, Eb
- PLExpCutOff: Prefactor, Index, Scale, Cutoff
-

```

1 crab:
2   name: crab
3   spatialModel:
4     DEC:
5       free: 0
6       max: 24
7       min: 20
8       name: DEC
9       scale: 1
10      value: 22.02
11    RA:
12      free: 0
13      max: 84.0
14      min: 83.0
15      name: RA
16      scale: 1
17      value: 83.63
18    type: SkyDirFunction
19  spectrum:
20    Index:
21      error: 0.036741536956782284
22      free: true
23      max: 10
24      min: 0
25      name: Index
26      scale: -1
27      value: 3.0526707697089317
28    Prefactor:
29      error: 0.02858086637415047
30      free: true
31      max: 100
32      min: 0
33      name: Prefactor
34      scale: 1e-14
35      value: 1.059534645016219
36    Scale:
37      free: 0
38      max: 100
39      min: 0
40      name: Scale
41      scale: 1
42      value: 20
43    type: PowerLaw

```

```

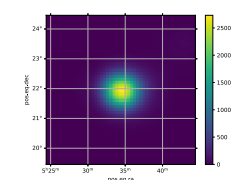
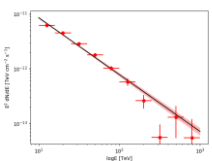
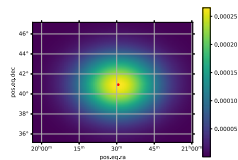
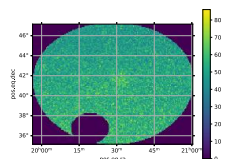
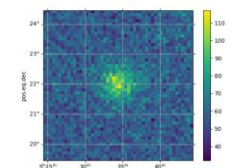
1 crab:
2   name: crab
3   type: PointSource
4   spatialModel:
5     type: SkyDirFunction
6   RA: ...
7   DEC: ...
8   spectrum:
9     type: LogParabola
10  norm: ...
11  alpha: ...
12  beta: ...
13  Eb: ...
14
15 newsrc:
16  name: newsrc
17  type: PointSource
18  spatialModel:
19    type: SkyDirFunction
20  RA: ...
21  DEC: ...
22  spectrum:
23    type: PowerLaw
24  Prefactor: ...
25  Index: ...
26  Scale: ...

```


For Users - Output

Scripts:

- DataSelect.py name ra dec dx dy
- plotCountsmap.py noncube.fits/nbkgcube.fits
- plotfitRegion.py fitRegion.yaml
- plotSpatialMap.py model_input.yaml srcName
- plotSpectrum.py model_input.yaml srcName
- BinnedAnalysis.py srcName logEmin logEmax
 - plotSED.py srcName logEmin logEmax Ebins
 - plotTSmap.py TSmap.fits



```

1  inRegion:
2    CygnusCocoon:
3      xref:
4        value: 307.17
5        unit: degree
6      yref:
7        value: 41.17
8        unit: degree
9      frame: fk5
10     rad:
11       value: 6
12       unit: degree
13
14  outRegion:
15     src:
16       xref:
17         value: 304.85
18         unit: degree
19       yref:
20         value: 36.80
21         unit: degree
22       frame: fk5
23       rad:
24         value: 1.5
25         unit: degree
  
```

```

1  LikelihoodValue: -708945.8533701402
2  Nbkg_LowerError: '-0.003'
3  Nbkg_UpperError: '0.003'
4  Nbkg_error: '0.003'
5  Nbkg_value: '1.006'
6  crab:
7    DEC_value: '22.02'
8    Flux_error: '1.31e-14'
9    Flux_value: '4.19e-13'
10   Index_LowerError: '-0.04'
11   Index_UpperError: '0.04'
12   Index_error: '0.04'
13   Index_scale: '-1.00e+00'
14   Index_value: '3.04'
15   Npred: '4928.40'
16   Prefactor_LowerError: '-0.03'
17   Prefactor_UpperError: '0.03'
18   Prefactor_error: '0.03'
19   Prefactor_scale: '1.00e-14'
20   Prefactor_value: '1.04'
21   RA_value: '83.63'
22   Scale_value: '20.00'
23   TsValue: 2830.3610210744664
  
```

For Developers

Data: on/off cube data

Time: lt_mjd or lt_lst

IRFs: a_{eff} , $psfcube$, $ediscube$

More SpatialModel and Spectrum

```
class PowerLaw(Spectrum):
    def GetSpecFunc(self, e):
        n = self.GetParValue('Prefactor')
        gamma = self.GetParValue('Index')
        e0 = self.GetParValue('Scale')
        return n * (e / e0)**gamma

class BrokenPowerLaw(Spectrum):
    def GetSpecFunc(self, e):
        n = self.GetParValue('Prefactor')
        gamma1 = self.GetParValue('Index1')
        gamma2 = self.GetParValue('Index2')
        eb = self.GetParValue('BreakValue')

        if type(e) in (int, float):
            if e < eb:
                dnde = n * (e / eb)**gamma1
            else:
                dnde = n * (e / eb)**gamma2
        else:
            dnde = np.zeros_like(e)
            dnde[e <= eb] = n * (e[e <= eb] / eb)**gamma1
            dnde[e >= eb] = n * (e[e >= eb] / eb)**gamma2
        return dnde
```

```
class RadialGaussian(SpatialModel):
    def GetSpatialMap(self, ccube=None):
        self.RA = self.GetParValue('RA')
        self.DEC = self.GetParValue('DEC')
        self.Sigma = self.GetParValue('Sigma')

        if ccube:
            skycrd = self.GetSkycrd()
            sep = skycrd.separation(ccube.skycrd).to(u.rad).value
            data = 1/(2*np.pi*np.deg2rad(self.Sigma)**2) * np.exp(-sep**2 / np.deg2rad(self.Sigma)**2 / 2).reshape(ccube.nxpix, ccube.nypix)
            wcs = ccube.wcs
        else:
            deltax = deltay = 0.1
            nxpix = nypix = 10 * self.Sigma / deltax
            sigma = self.Sigma / deltax
            mod = models.Gaussian2D(1., nxpix/2, nypix/2, sigma, sigma, 0)

            x, y = np.mgrid[0:nxpix, 0:nypix]
            data = mod(x, y)
            data = data / (data * np.deg2rad(deltax) * np.deg2rad(deltay)).sum()
            wcs = mapproj.make_wcs(nxpix, nypix, deltax, deltay, 'C', self.RA, self.DEC, 'CAR')
        return SkyMap.NewSpatialMap(data, wcs)

    def GetSkycrd(self):
        return SkyCoord(ra=self.RA*u.degree, dec=self.DEC*u.degree, frame='fk5')
```

Any suggestions or contributions are welcome!



Summary and Plans

We provide a 3D likelihood analysis tool for KM2A data analysis.

With the half-array data and IRFs, the results are consistent with traditional method.

We plan to apply this tool for 3-quarters and full-array of KM2A and WCDA data.

Summed likelihood for KM2A data and joint Fit for KM2A and WCDA data.

Append Healpix projection for large region analysis.



Thanks for your attention