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GalSim Quick Reference

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1. Overview

The GalSim package provides a number of Python classes and methods for simulating astronomical images. We assume GalSim is installed; see the GalSim Wiki¹ or the file INSTALL.md in the base directory for instructions. The package is imported into Python with

```
>>> import galsim
```

and the typical work flow, as demonstrated in the example scripts in the examples / directory (all paths given relative to the GalSim base directory from now on), will normally be something like the following:

- Construct a representation of your desired astronomical object as an instance of either the GSObject or ChromaticObject class, which represent (possibly wavelength-dependent) surface brightness profiles (of galaxies or PSFs). Multiple components can be combined using the special Add and Convolve functions see Section 2.
- Apply transformations such as shears, shifts, rotation, or magnification using the methods of the GSObject or ChromaticObject see Sections 2.4 & 3.
- o Draw the object into a GalSim Image, representing a postage stamp image of your astronomical object. This can be done using the obj.drawImage(...) method carried by all GSObjects and ChromaticObjects for rendering images see Sections 2.4 & 5.
- Add noise to the Image using one of the GalSim random deviate classes see Section 4.
- Add the postage stamp Image to a subsection of a larger Image instance see Section 5.2.
- Save the Image(s) to file in FITS (Flexible Image Transport System) format see Sections 5.2 & 6.5.

There are many examples of this workflow in the directory examples/, showing most of the GalSim library in action, in the scripts named demol.py - demol3.py. An online tutorial guide outlines the capabilities that are demonstrated in the demos. This document provides a brief, reference description of the GalSim classes and methods which can be used in these workflows.

Where possible in the following Sections this document has been hyperlinked to the online GalSim documentation generated by *doxygen*, where a more detailed description can be found. We also suggest accessing the full docstrings for *all* the classes and functions described below in Python itself, e.g. by typing

```
>>> help(galsim.<ObjectName>)
```

within the Python interpreter. If using the ipython package, which is recommended, instead simply type

```
In [1]: galsim.<ObjectName>?
```

and be sure to use the excellent tab-completion feature to explore the many methods and attributes of the GalSim classes.

¹ Note: links to various web pages are colored blue in this document to help you find more information.

2. GSObjects

2.1. GSObject classes and when to use them

GalSim has many classes that represent various types of surface brightness profiles. These are all subclasses of the GSObject base class, which encapsulates much of the functionality that is common to all such profiles in GalSim. All but the last two classes listed below are 'simple' GSObjects that can be initialized by providing values for their required and optional parameters. The last two are 'compound' classes used to represent combinations of GSObjects.

They are summarized in the following hyperlinked list, in which we also give the required parameters for initializing each class in parentheses after the class name. For more information and initialization details for each GSObject, the Python docstring for each class is available within the Python interpreter, for example for Sersic the documentation would be accessed using

```
>>> help(galsim.Sersic)
```

Alternatively follow the hyperlinks on the class names listed below to view the documentation based on the Python docstrings.

In cases where multiple options for specifying the object size exist we list these in the object description. We also show some of the optional parameters available for use (e.g. total flux) along with default values:

```
o galsim.Gaussian(size, flux=1.)
```

A circular 2D Gaussian surface brightness profile. Requires one of the following size parameters: sigma, fwhm, or half_light_radius.

```
o galsim.Moffat(beta, size, flux=1.)
```

A Moffat profile with slope parameter beta, used to approximate ground-based telescope PSFs. Requires one of the following <code>size</code> parameters: <code>scale_radius</code>, <code>fwhm</code>, or <code>half_light_radius</code>. For information about other optional parameters, see the documentation for this object.

```
o galsim.Airy(lam_over_diam, obscuration=0., flux=1.)
```

An Airy PSF for ideal diffraction through a circular aperture, parametrized by the wavelength-aperture diameter ratio lam_over_diam, with optional obscuration.

```
o galsim.Kolmogorov(size, flux=1.)
```

A Kolmogorov PSF for long-exposure images through a turbulent atmosphere. Requires one of the following size parameters: lam_over_r0, fwhm, or half_light_radius.

```
o galsim.OpticalPSF(lam_over_diam, flux=1.)
```

A simple model for non-ideal (aberrated) propagation through circular/square apertures, parametrized by the wavelength- aperture dimension ratio lam_over_diam, with optional obscuration. For information about other optional parameters, see the documentation for this object.

```
o galsim.InterpolatedImage(image, ...)
```

A surface brightness profile for which we have an Image representation, which is interpolated appropriately to locations other than the pixel centers. For information about other optional parameters, see the documentation for this object.

```
o galsim.Pixel(scale, flux=1.)
  A square tophat profile, used for representing pixels with pixel scale scale.
o galsim.Box(width, height, flux=1.)
  An arbitrary rectangular box profile.
o galsim.TopHat(radius, flux=1.)
  A circular tophat profile.
o galsim.Sersic(n, half_light_radius, flux=1.)
  A profile form the Sérsic family of galaxy light profiles, parametrized by an index n and half_light_radius.
o galsim.Exponential(size, flux=1.)
  An Exponential galaxy disc profile, a Sérsic with index n=1. Requires one of the following size parameters:
  scale_radius or half_light_radius.
o galsim.InclinedExponential(i, scale_radius, scale_height, flux=1.)
  An Inclined Exponential galaxy disc profile. Requires the following parameters: i, scale_radius, scale_height
o galsim.DeVaucouleurs(half_light_radius, flux=1.)
  A De Vaucouleurs galaxy bulge profile, a Sérsic with index n=4 and input half_light_radius.
o galsim.Spergel(nu, size, flux=1.)
  A class describing a Spergel (2010) profile. Requires one of the following size parameters: half_light_radius
  or scale_radius.
o galsim.RealGalaxy(real_galaxy_catalog, ...)
  A galaxy model taken from an image of a real galaxy and includes a correction for the original PSF. Use of
  this class requires the download of external data. RealGalaxy objects can be directly instantiated based on a
  catalog that is input via the real_galaxy_catalog parameter (an instance of the RealGalaxyCatalog
  class), or using the galsim. COSMOSCatalog class.
```

An example catalog of 100 real galaxies is in the repository itself; a set of real galaxy images, with original PSFs, can be downloaded following instructions on the *RealGalaxy Data Download Page* on the GalSim Wiki, or using an executable galsim_download_cosmos that is distributed with GalSim. For information about optional parameters, see the documentation for this object.

```
    galsim.Sum([list of objects])
    A compound object representing the sum of multiple GSObjects.
    galsim.Convolution([list of objects])
    A compound object representing the convolution of multiple GSObjects.
```

Note that the last two objects, Sum and Convolution, are usually created by invoking the Add and Convolve functions. These functions will automatically create ChromaticSum and ChromaticConvolution objects instead if any of their arguments are ChromaticObjects instead of GSObects (see Section 3).

Also note that all of the GSObjects except for RealGalaxy, Sum, and Convolution require the specification of one size parameter.

2.2. Atmosphere

GalSim can also simulate PSFs generated by turbulent layers in the atmosphere over a wide field of view. (Though note that this machinery is much slower than other parts of GalSim.) The syntax for creating this type of GSObject is a little bit different than for those above. First, some number AtmosphericScreens should be assembled into a PhaseScreenList:

```
>>> r0_500 = 0.2 # Fried parameter (at 500 nm) in meters
>>> screen_size = 100 # Size of atmospheric screen in meters.
>>> layer0 = galsim.AtmosphericScreen(screen_size=screen_size, r0_500=r0_500)
>>> layer1 = ...
>>> psl = galsim.PhaseScreenList([layer0, layer1, layer2])
```

Note that AtmosphericScreens have parameters for altitude (which affects how PSFs at different field angles are related) and wind speed (which lets the screens evolve in time over an exposure), in addition to parameters like r0_500 that control the fixed pattern of turbulence in a screen.

To make a PSF, a telescope aperture is also needed:

```
>>> diam = 4.0 # Meters
>>> obscuration = 0.5 # Linear fractional obscuration
>>> aper = galsim.Aperture(diam=diam, obscuration=obscuration)
```

Then to make PSFs, use the PhaseScreenList.makePSF method, optionally specifying a field angle (i.e., where on the focal plane to generate the PSF):

```
>>> exptime = 30.0 # Exposure time in seconds.
>>> lam = 700.0 # Wavelength in nanometers.
>>> psf0 = psl.makePSF(exptime=exptime, lam=lam, aper=aper)
>>> psf1 = psl.makePSF(exptime=exptime, lam=lam, aper=aper,
... theta_x=0.1*galsim.arcmin, theta_y=0.1*galsim.arcmin)
```

One caveat is that when calling makePSF sequentially, as above, the atmospheric screens cover different time intervals for the different calls (e.g., starting at 0 seconds and starting at 30 seconds for the two calls above). To generater two PSFs at different field angles but for the *same* 30 second interval, pass theta_x and theta_y as lists to makePSF:

2.3. Units

The choice of units for the size parameters is up to the user, but it must be kept consistent between all GSObjects. These units must also adopted when specifying the Image pixel scale, whether this is set via the GSObject instance method obj.drawImage(...) (see Section 2.4), or when constructing the Image (see Section 5).

As an example, consider the lam_over_diam parameter, which provides an angular scale for the Airy via the ratio λ/D for light at wavelength λ passing through a telescope of diameter D. Putting both λ and D in metres and taking the ratio gives lam_over_diam in radians, but this is not a commonly used angular scale when describing astronomical objects such as galaxies and stellar PSFs, nor is it often used for image pixel scales. If wishing to use arcsec, which is more common in both cases, the user should multiply the result in radians by the conversion factor $648000/\pi$. In principle, however, any consistent system of units could be used. Users wishing to separately specify the values of λ and D with their natural units of nanometers and meters may do so using lam and diam; in that case, the keyword scale_unit can be used to define the system of units used for the scales in images.

2.4. Important GSObject methods

A number of methods are shared by all the GSObjects of Section 2. In what follows, we assume that a GSObject labelled obj has been instantiated using one of the calls described in the documentation linked above. For example,

```
>>> obj = galsim.Sersic(n=3.5, half_light_radius=1.743).
```

Once again, for more information regarding each galsim. GSObject method, the Python docstring is available

```
>>> help(obj.<methodName>)
```

within the Python interpreter. Alternatively follow the hyperlinks on the class names above to view the documentation based on the Python docstrings.

Some of the most important and commonly-used methods for such an instance are:

```
obj.centroid()
Return the (x, y) centroid of the GSObject as a PositionD (see Section 6.2).
obj.getFlux()
Get the flux of the GSObject.
obj.withScaledFlux(flux_ratio)
Return a version of the GSObject with the flux scaled by flux_ratio.
obj.withFlux(flux)
Return a version of the GSObject with the flux set to flux.
obj.dilate(scale)
Return a version of this GSObject with the linear size dilated by a factor scale, conserving flux.
obj.magnify(mu)
Return a version of this GSObject with the area magnified by a factor mu, conserving surface brightness.
```

```
o obj.shear(...)
```

Return a version of this GSObject that has been sheared by some amount. This method can handle a number of different input conventions for the shear (see also Shear; Section 6.3). Commonly-used input conventions (supplied as keyword arguments, default values zero):

- obj.shear(g1=g1, g2=g2)
 - Apply the first (g1) and second (g2) component of a shear defined so that |g| = (a b)/(a + b) where a and b are the semi-major and semi-minor axes of an ellipse.
- obj.shear(e1=e1, e2=e2)
 - Apply the first (e1) and second (e2) component of a shear defined so that $|e| = (a^2 b^2)/(a^2 + b^2)$ where a and b are the semi-major and semi-minor axes of an ellipse.
- obj.shear (g=g, beta=beta) Apply magnitude (g) and polar angle (beta) of a shear defined using the |g| definition above.
- obj.shear(e=e, beta=beta) Apply magnitude (e) and polar angle (beta) of a shear defined using the |e| definition above.
- o obj.rotate(theta)

Return a version of the GSObject that has been rotated by an angle theta (positive direction anti-clockwise), where theta is an Angle instance (see Section 6.1).

o obj.shift(dx, dy)

Return a version of the GSObject with its centroid position shifted by (dx, dy).

Draw and return an Image (see Section 5) of the GSObject. Some information about important optional parameters (see the linked Python docstrings for more detail), along with default values:

- image (default = None)
 - If supplied, the drawing will be done into a user-supplied Image instance image. If not supplied (i.e. image = None), an automatically-sized Image instance will be returned.
- scale (default = None)

The optional image pixel scale, which if provided should use the same units as used for the GSObject size parameters.

- wcs (default = None)

The wcs may optionally be provided in lieu of a simple pixel scale, in which case this would specify the conversion between image coordinates and world (aka sky) coordinates. The GSObject is taken to be defined in world coordinates and this function tells GalSim how to convert to image coordinates when it draws the profile. If neither scale nor wcs are provided here, then GalSim will use the wcs attribute of the image if available. Otherwise, it will use the Nyquist scale given the maximum modeled frequency in the GSObject.

- method (default = 'auto')
 - * 'auto' chooses either 'fft' or 'real_space' appropriately based on the kind of profile being drawn.
 - * 'fft' renders the image using a discrete Fourier transform to convolve by the pixel response.

- * 'real_space' uses direct integration to integrate the flux over the pixels if possible. (It defaults to 'fft' when such a procedure is impossible or impractical.)
- * 'phot' renders the image by shooting a finite number of photons. The resulting rendering therefore contains stochastic noise, but uses few approximations. Note however, that you cannot use 'phot' with a RealGalaxy instance. This method has a few additional optional parmaeters; see below.
- * 'no_pixel' does not integrate over the pixel response, but rather samples the profile directly at the pixel centers and multiplies by the pixel area. This is the appropriate choice if your PSF already includes the convolution by the pixel response, e.g. if it comes from an image of a star observed on the same camera.
- * 'sb' is simlar to 'no_pixel' except that it does not scale the values by the pixel area. So the drawn values will be direct samples of the surface brightness at each location.
- add_to_image (default = False)
 Whether to add flux to a supplied image rather than clear out anything in the image before drawing.

If method = 'phot', there are a number of additional optional parameters. Important examples worthy of mention are:

- n_photons (default = 0)
 If provided, the number of photons to use. If not provided, use as many photons as necessary to end up with an image with the correct poisson shot noise for the object's flux. (Normally, this means use n_photons=flux, since flux is taken to be in units of photons, but there are exceptions to this.)
- max_extra_noise (default = 0.)
 If provided, the allowed extra noise in each pixel. This is only relevant if n_photons = 0, so the number of photons is being automatically calculated. In that case, if the image noise is dominated by the sky background, you can get away with using fewer shot photons than the full n_photons = flux. Essentially each shot photon can have a flux > 1, which increases the noise in each pixel. The max_extra_noise parameter specifies how much extra noise per pixel is allowed because of this approximation.
- poisson_flux (default = True)
 Whether to allow total object flux scaling to vary according to Poisson statistics for n_photons samples.

The drawImage method has a number of additional optional parameters. Please see the linked Python docstrings for more details.

Finally, you may see by exploring the docstrings that many of the GSObject instances also have their own specialized methods, often for retrieving parameter values. Examples are obj.getSigma() for the Gaussian, or obj.getHalfLightRadius() for many of the GSObjects.

3. Chromaticity

Wavelength-dependent surface brightness profiles are represented as galsim.ChromaticObject instances in GalSim. These objects generally require an galsim.SED to be created, and always require a galsim.Bandpass object in order to draw. Thus we will go over SEDs and Bandpasses first.

3.1. Bandpasses

The galsim.Bandpass class represents a spectral throughput function, which could be an entire imaging system throughput response function (reflection off of mirrors, transmission through filters, lenses and the atmosphere, quantum efficiency of detectors), or individual pieces thereof. Bandpasses, together with spectral energy distributions (SEDs; below) are necessary to compute the relative contribution of each wavelength of a ChromaticObject to a drawn image.

Bandpass instances may be initialized in several ways:

- o galsim.Bandpass(filename, wave_type, ...) where filename points to a text file with two columns, the first for wavelength and the second for dimensionless throughput. wave_type is used to tell GalSim whether the wavelengths in that file are in units of nanometers or Angstroms.
- o galsim.Bandpass (function, wave_type, red_limit, blue_limit, ...) where function is a python function that accepts wavelength and returns dimensionless throughput, and wave_type is as above. The keywords red_limit and blue_limit are required in this case to specify the integration limits of the bandpass.

For Bandpass instances initialized from a file, the following two methods can be used to reduce the number of samples used for integrations (and hence reduce the time it takes to draw a ChromaticObject).

- bandpass.truncate(blue_limit, red_limit, relative_throughput, ...)
 Return a new bandpass truncated to only include wavelengths between blue_limit and red_limit, or truncated based on the supplied relative_throughput level.
- o bandpass.thin(rel_err,...)
 Return a thinner version of the bandpass by removing sample values (locations where $F(\lambda)$ is defined) while retaining the accuracy of the integral over the bandpass to the stated relative error rel_err. Other aspects of the behavior of this routine are set via additional keyword arguments.

Finally, note that Bandpass instances may be multiplied together and are callable, returning dimensionless throughput as a function of wavelength in nanometers.

3.2. SEDs

Spectral energy distributions, respresented by the SED class, may be constructed in several ways, similarly to band-passes:

o galsim.SED(filename, wave_type, flux_type, ...) where filename points to a text file with two columns, the first for wavelength in units given by wave_type and the second for flux density in units given by flux_type. o galsim.SED (function, wave_type, flux_type, ...) where function is a python function that accepts wavelength in units given by wave_type and returns flux density in units given by flux_type.

Important methods for SED objects include:

- sed.withFluxDensity(target_flux_density, base_wavelength)
 Return a new SED with the same functional shape, but with the flux density (in units of ergs/nm) at the reference wavelength base_wavelength set to target_flux_density. Note that SED objects are immutable, so the original SED is unchanged.
- sed.calculateFlux (bandpass)

 Calculate and return the flux transmitted through a bandpass in photons.
- sed.withFlux(target_flux, bandpass)
 Return a new SED for which the flux transmitted through the input bandpass is target_flux.
- o sed.atRedshift(z)

Return a new SED instance with wavelength shifted be at redshift z. Note that SED instances remember their redshifts (except when created as sums and differences of other SEDs), so applying this method a second twice with the same argument z is equivalent to applying it just once.

Finally, note that SED instances can be added together, multiplied by scalars or functions (of wavelength in nanometers), and are callable, returning flux density in photons/nm.

3.3. ChromaticObjects

Chromatic surface brightness profiles are generally constructed by modifying an existing GSObject. The simplest possible ChromaticObject can be formed by passing a GSObject to the ChromaticObject constructor:

```
>>> obj = galsim.Gaussian(fwhm=1.0)
>>> chromatic_obj = galsim.ChromaticObject(obj)
```

At this stage, chromatic_obj represents the same profile as obj, but now has access to ChromaticObject methods.

The simplest way to construct a non-trivial chromatic object is to multiply a GSObject by an SED. This creates a separable wavelength-dependent surface brightness profile $I(x, y, \lambda) = I_0(x, y) f(\lambda)$:

```
>>> chromatic_obj = obj * sed
```

ChromaticObjects may be combined and transformed similarly to GSObjects, using the functions and methods Add, Convolve, withScaledFlux, dilate, magnify, shear, rotate, and shift.

The dilate and shift methods of ChromaticObjects can also accept as an argument a function of wavelength (in nanometers) that returns a wavelength-dependent dilation or shift. These can be used to implement chromatic PSFs. For example, a diffraction limited PSF might look like:

```
>>> psf500 = galsim.Airy(lam_over_diam=2.0)
>>> chromatic_psf = ChromaticObject(psf500).dilate(lambda w:(w/500)**(1.0))
```

The drawImage method of a ChromaticObject is similar to the drawImage method of a GSObject, except that it requires a Bandpass object as its first argument.

```
>>> gband = galsim.Bandpass(lambda w:1.0, wave_type='nm', blue_limit=410, red_limit=550)
>>> image = chromatic_obj.drawImage(gband)
```

GalSim comes with built-in support for ground-based PSFs affected by differential chromatic refraction and Kolmogorov chromatic seeing (FWHM $\propto \lambda^{-0.2}$) through the following class:

o ChromaticAtmosphere (base_obj, base_wavelength, ...) where base_obj is the fiducial PSF at wavelength base_wavelength. There are a number of options for specifying details like the position of the object with respect to zenith and the physical properties of the atmosphere.

There are also classes for representing chromatic optical PSFs:

o ChromaticAiry(...)

for representing a chromatic Airy profile, where the user must specify either lam (the telescope diameter in meters) or lam_over_diam (the ratio of some fiducial wavelength to the diameter).

• ChromaticOpticalPSF (...) for representing a fully chromatic optical PSF, including chromatic aberrations.

For sufficiently complex chromatic objects, an optional optimization can be invoked to speed up the image rendering process; for more information, view the docstring for the setupInterpolation method associated with ChromaticObjects.

4. Random deviates

4.1. Random deviate classes and when to use them

Random deviates can be used to add a stochastic component to the modeling of astronomical images, such as drawing object parameters according to a given distribution or generating random numbers to be added to image pixel values to model noise.

We now give a short summary of the 8 random deviates currently implemented in GalSim. The optional parameter seed listed below is used to seed the pseudo-random number generator: it can either be omitted (the random deviate seed will be set using the current time), set to an integer seed, or used to pass another random deviate (the new instance will then use and update the same underlying generator as the input deviate).

The deviates, with a description of their distributions, parametrization and default parameter values, are as follows:²

² Note that, with the exception of DistDeviate, all the deviate classes are defined in the C++ layer, so the hyper-links below link to that code, rather than the Python documentation. The syntax is not too different in Python, but unfortunately, the docstrings are not currently linkable directly. For more information, please refer to the full docstrings, accessible via commands like help(galsim.UniformDeviate) in the Python interpreter.

```
o galsim.UniformDeviate(seed)
  Uniform distribution in the interval [0, 1).
o galsim.GaussianDeviate(seed, mean=0., sigma=1.)
  Gaussian distribution with mean and standard deviation sigma.
o galsim.BinomialDeviate(seed, N=1, p=0.5)
  Binomial distribution for N trials each of probability p.
o galsim.PoissonDeviate(seed, mean=1.)
  Poisson distribution with a single mean rate.
o galsim.WeibullDeviate(seed, a=1., b=1.)
  Weibull distribution family (includes Rayleigh and Exponential) with shape parameters a and b.
o galsim.GammaDeviate(seed, alpha=1., beta=1.)
  Gamma distribution with parameters alpha and beta.
o galsim.Chi2Deviate(seed, n=1.)
  \chi^2 distribution with degrees-of-freedom parameter n.
o galsim.DistDeviate(seed, function, x_min, x_max)
  Use an arbitrary function for P(x) from x_min .. x_max.
```

It is possible to specify the random seed so as to get fully deterministic behavior of the noise when running a particular script.

4.2. Important random deviate methods

We now illustrate the most commonly-used methods of the random deviates, assuming that some random deviate instance dev has been instantiated, for example by

```
>>> dev = galsim.GaussianDeviate(sigma=3.9, mean=50.).
```

The most important and commonly-used method for such instances is:

o dev()

Calling the deviate directly simply returns a single new random number drawn from the distribution represented by dev. As an example:

```
>>> dev = galsim.UniformDeviate(12345)
>>> dev()
0.9296160866506398
>>> dev()
0.8901547130662948
```

4.3. Noise models

One common way to use the random deviates is as part of a noise model for adding noise to an image. These have their own separate hierarchy of classes

o galsim.GaussianNoise(rng, sigma=1.)

Every pixel gets Gaussian noise with rms sigma, using the random number generator given as rng, which may be any deviate instance.

o galsim.PoissonNoise(rng, sky_level=0.)

Every pixel gets Poisson noise according to the flux in the image plus an option sky level, sky_level, using the random number generator given as rng, which may be any deviate instance.

o galsim.DeviateNoise(dev)

The noise value for every pixel is drawn from the given Deviate instance dev.

o galsim.CCDNoise(rng, sky_level=0., gain=1., read_noise=0.)

A combination of Poisson noise (with a gain value in electrons/ADU) and Gaussian read noise, using the random number generator given as rng, which may be any deviate instance.

o galsim.VariableGaussianNoise(rng, var_image)

Every pixel gets Gaussian noise with variance given by the var_image parameter (an Image instance), using the random number generator given as rng, which may be any deviate instance. The supplied var_image must be the same size and shape as the Image onto which the noise is eventually added.

o galsim.CorrelatedNoise(image, rng, ...)

The provided image should be a pure noise field (i.e. a blank area of sky with no detectable objects). The correlated noise of this image will be computed, which can then be applied to any other image, laying down noise with the same correlations. The provided random number generator, rng, which may be any deviate instance will be used as the basis of the required random numbers..

To apply noise to an Image using these noise models, the command is simply:

o image.addNoise(noise)

This adds stochastic noise, according to the noise model noise, to each element of the data array in the Image instance image.

In addition, GalSim can remove some or all of the noise correlations in an image that has a known correlated noise:

o image.whitenNoise(noise)

This adds extra correlated noise to the image such that if the image initially has correlated noise given by noise (which should be a CorrelatedNoise object), then the image will end up with white noise. It tries to add the minimum amount of extra noise possible to achieve this.

o image.symmetrizeNoise(noise, order=4)

This is similar to whitenNoise, except that rather than making the resulting noise profile completely white, it just makes the final noise in the image have N-fold symmetry where N is given by order. This typically requires much less noise to be added than whitenNoise, and for many purposes this is sufficient to remove the adverse effects of correlated noise.

5. Images

5.1. Image classes and when to use them

The GalSim Image class stores array data, along with the bounds of the array, and a function that converts between image coordinates and world coordinates (also known as sky coordinates). The most common World Coordinate System (WCS) function that you will encounter is a simple scaling of the units from pixels to arcsec. This WCS can be specified simply by im.scale, as we have seen already. More complicated WCS functions would need to be referenced via im.wcs. See the docstring for BaseWCS for more details.

Image instances can be operated upon to add stochastic noise simulating real astronomical images (see Section 4), and have methods for writing to FITS format output.

The most common way to initialize an image is with two integer parameters nx and ny, giving the image extent in the x and y dimensions, respectively. Example initialization is therefore:

```
o galsim. Image (nx, ny)
```

This would create an image with single precision (32 bit) floats for the data elements, which is usually the most appropriate type for astronomical images. However, you can specify other types for the data using a suffix letter after Image:

```
    galsim.ImageS(nx, ny) for 16 bit integers.
    galsim.ImageI(nx, ny) for 32 bit integers.
    galsim.ImageF(nx, ny) for single precision (32 bit) floats.
    galsim.ImageD(nx, ny) for double precision (64 bit) floats.
```

Other ways to construct an Image can be found in the docstrings, including via an input NumPy array.

To access the data as a NumPy array, simply use the image.array attribute, where image is an instance of one of these Image classes. However, note that the individual elements in the array attribute are accessed as image.array[y,x], matching the standard NumPy convention, while the Image class's own accessors are all (x,y) in ordering.

5.2. Important Image methods and operations

We now illustrate the most commonly-used methods of Image class instances. We will assume that some Image instance image has been instantiated, for example by

```
>>> image = galsim.Image(100, 100).
```

This Image instance is then ready to pass to a GSObject for drawing. The most important and commonly-used methods for such an instance are:

- o image.bounds
 - Get the bounding box of the data.
- o image.wcs

Get or set the WCS function to convert between image coordinates and world coordinates.

o image.scale

Get or set the pixel scale scale for this image. The getter only works if the WCS is really just a pixel scale. The setter will make it a pixel scale.

o image.addNoise(noise)

This adds stochastic noise according to the given noise model, noise to each element of the data array in image. This is the method previously referenced in Section 4.

o image.write(file_name, ...)

Write the image to a FITS file, given by file_name. There are other options for how to write this (such as writing to an HDUList rather than a file), so see the docstring for more details, including details about the optional parameters. In Section 6.5 we discuss how to write to multi-extension FITS files.

Image instances are also returned when accessing a sub-section of an existing Image. For example

```
>>> imsub = image.subImage(bounds)
```

where bounds is a Bounds I instance (see Section 6.2) assigns imsub as an view into the sub-region of image lying in the area represented by bounds. Equivalent syntax is also

```
>>> imsub = image[bounds]
```

It is also possible to change the values of a sub-region of an image this way, for example

```
>>> image[imsub.bounds] += imsub
```

if wishing to add the contents of imsub to the area lying within its bounds in image. Note that here we have made use of the image. bounds attribute carried by all of the Image classes.

6. Miscellaneous classes and functions

A summary of miscellaneous GalSim library objects, subcategorized into broad themes. As ever, docstrings for *all* the classes and functions below can be accessed via help (galsim.<Name>) within the Python interpreter.

6.1. Angles

o galsim.Angle(value, angle_unit)

Class to represent angles (with multiple unit types), which can be initialized by multiplying a numerical value and an AngleUnit instance angle_unit (see below).

o galsim.AngleUnit

There are five built-in AngleUnits which are always available for use:

```
- galsim.radians
```

- galsim.degrees
- galsim.hours
- galsim.arcmin
- galsim.arcsec

Please see the Python docstrings for information about defining your own AngleUnits.

6.2. Bounds and Positions

```
o galsim.BoundsI(...)
galsim.BoundsD(...)
```

Classes to represent image boundaries as the vertices of a rectangle.

```
o galsim.PositionI(x, y)
galsim.PositionD(x, y)
```

Classes to represent 2D positions on the x-y plane, e.g., for describing object centroid positions.

For both bounds and positions, the I and D refer to integer and double-precision floating point representations.

6.3. Shears

```
o galsim.Shear(...)
```

Class to represent shears in a variety of ways. This class can be initialized using a variety of different parameter conventions (see the doc string for the complete list). Commonly-used examples (supplied as keyword arguments, default values zero):

- galsim. Shear (g1=g1, g2=g2) set via the first (g1) and second (g2) component of a shear defined so that |g| = (a-b)/(a+b) where a and b are the semi-major and semi-minor axes of an ellipse.
- galsim. Shear (e1=e1, e2=e2) set via the first (e1) and second (e2) component of a shear defined so that $|e|=(a^2-b^2)/(a^2+b^2)$ where a and b are the semi-major and semi-minor axes of an ellipse.
- galsim.Shear (g=g, beta=beta) set via magnitude (g) and polar angle (beta) of a shear defined according to the |g| definition above.
- galsim. Shear (e=e, beta=beta) set via magnitude (e) and polar angle (beta) of a shear defined according to the |e| definition above.

6.4. Lensing shear fields

GalSim has functionality to simulate scientifically-motivated lensing shear fields. The two relevant classes for users are:

o galsim.PowerSpectrum(...)

Represents a flat-sky shear power spectrum P(k), where the E and B-mode power spectra can be separately specified as E-power_function and B-power_function. The getShear(...) method is used to generate a random realization of a shear field from a given PowerSpectrum object, and there are methods to get convergence or magnification as well.

o galsim.NFWHalo(...)

Represents a matter density profile corresponding to a projected, circularly-symmetric NFW profile such as might be used to simulate lensing by a galaxy cluster. This class has two methods of interest for users, getShear() and getConvergence(), which can be used to get the shears and convergences at *any* (nongridded) image-plane position.

These classes have additional requirements on the units used to specify positions; see the documentation for these classes for more details.

The GalSim repository also contains a module with a PowerSpectrumEstimator class that can be used to estimate shear power spectra from gridded shear values. See the linked documentation for more information.

6.5. Additional FITS input/output tools

For all of these functions, there are also options for reading/writing directly from/to an HDUList rather than a file, so see the linked doc strings for more details. Optional parameters control how to handle compression, although the default in all cases is to try to use the correct compression scheme based on the file's extension.

- image = galsim.fits.read(file_name, ...)
 Returns an Image instance image from a FITS file, file_name. If the FITS file has a WCS defined in the header, then GalSim will attempt to read that WCS and store it as image.wcs.
- o image_list = galsim.fits.readMulti(file_name, ...)
 Returns a Python list of Image instances (image_list) from a Multi-Extension FITS file, file_name.
- galsim.fits.writeMulti(image_list, file_name, ...)
 Write multiple Image instances stored in a Python list (image_list) to a Multi-Extension FITS file, file_name.
- galsim.fits.writeCube(image_list, file_name, ...)
 Write multiple Image instances stored in a Python list (image_list) to a three-dimensional FITS datacube, file_name.