First, let's start with the equation that you are fitting to. This is defined in the Imfit docs:

## GaussianModel

class GaussianModel(independent\_vars=['x'], prefix='', nan\_policy='raise', \*\*kwargs)

A model based on a Gaussian or normal distribution lineshape.

The model has three Parameters: amplitude, center, and sigma. In addition, parameters fwhm and height are included as constraints to report full width at half maximum and maximum peak height, respectively.

$$f(x;A,\mu,\sigma) = rac{A}{\sigma\sqrt{2\pi}}e^{[-(x-\mu)^2/2\sigma^2]}$$

where the parameter amplitude corresponds to A, center to  $\mu$ , and sigma to  $\sigma$ . The full width at half maximum is  $2\sigma\sqrt{2\ln 2}$ , approximately  $2.3548\sigma$ .

For more information, see: https://en.wikipedia.org/wiki/Normal\_distribution

- **Parameters: independent\_vars** (<u>list</u> of <u>str</u>, optional) Arguments to the model function that are independent variables default is ['x']).
  - prefix (<u>str</u>, optional) String to prepend to parameter names, needed to add two Models that have parameter names in common.
  - nan\_policy ({'raise', 'propagate', 'omit'}, optional) How to handle NaN and missing values in data.
     See Notes below.
  - \*\*kwargs (optional) Keyword arguments to pass to Model.

There is a sneaky line in here that is easy to miss, and that I think is the source of your confusion:

Parameters fwhm and height are included as constraints to report full width at half maximum and maximum peak height, respectively.

So, when you fit your data to the model you can extract both the height and amplitude parameters despite only the amplitude being optimised. This is because the two are related - to find out where height comes from, we can dive into the Imfit source code for the GaussianModel, and then from there piece together how height and amplitude are related.

```
class GaussianModel(Model):
   r"""A model based on a Gaussian or normal distribution lineshape.
   The model has three Parameters: 'amplitude', 'center', and 'sigma'.
   In addition, parameters 'fwhm' and 'height' are included as
   constraints to report full width at half maximum and maximum peak
   height, respectively.
   .. math::
       f(x; A, \mu, sigma) = \frac{A}{sigma}^2}}} e^{[{-\{(x-\mu)^2\}/{\{2 sigma}^2\}}]}
   where the parameter 'amplitude' corresponds to :math:'A', 'center' to
   :math:'\mu', and 'sigma' to :math:'\sigma'. The full width at half
   maximum is :math: 2\sigma\sqrt{2\ln{2}}, approximately
   :math: 2.3548\sigma .
   For more information, see: https://en.wikipedia.org/wiki/Normal_distribution
   ---
   fwhm_factor = 2*np.sqrt(2*np.log(2))
   height_factor = 1./np.sqrt(2*np.pi)
   def __init__(self, independent_vars=['x'], prefix='', nan_policy='raise',
       kwargs.update({'prefix': prefix, 'nan_policy': nan_policy,
                      'independent_vars': independent_vars})
       super().__init__(gaussian, **kwargs)
       self._set_paramhints_prefix()
   def _set_paramhints_prefix(self):
       self.set_param_hint('sigma', min=0)
       self.set_param_hint('fwhm', expr=fwhm_expr(self))
       self.set_param_hint('height', expr=height_expr(self))
   def guess(self, data, x, negative=False, **kwargs):
       """Estimate initial model parameter values from data."""
       pars = guess_from_peak(self, data, x, negative)
       return update_param_vals(pars, self.prefix, **kwargs)
def height_expr(model):
   """Return constraint expression for maximum peak height."""
   fmt = "{factor:.7f}*{prefix:s}amplitude/max({}, {prefix:s}sigma)"
   return fmt.format(tiny, factor=model.height_factor, prefix=model.prefix)
```

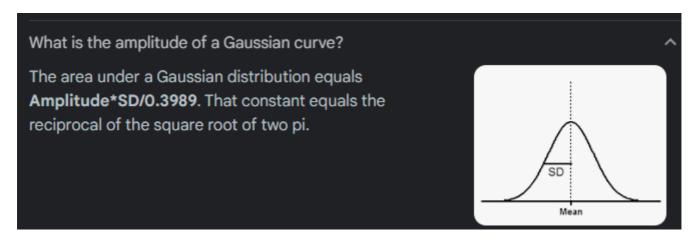
Combining these two things together leaves us with

```
height_factor = 1. / np.sqrt(2*np.pi) #recipricol of the square root of two pi
height = height_factor * amplitude / sigma
```

and rearranging to solve for amplitude gives

```
amplitude = height * sigma / height_factor
```

Now, if that looks vaguely familiar, that's because it is effectively the same as the area formula that you were trying to use!

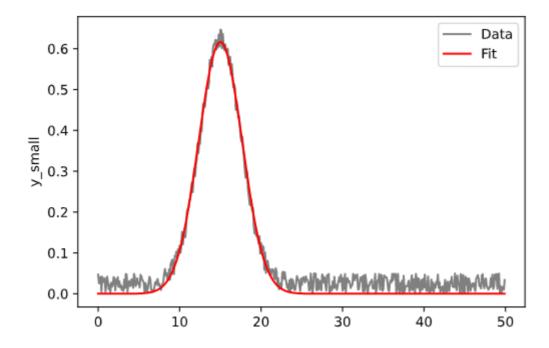


where, in this case SD=sigma.

So - long story short, the *amplitude* parameter that you are fitting using Imfit's gaussian model is in fact the area (composed of the sigma and height, and assuming the baseline is at 0). The reason you were getting strange results with your code is you were using this already-transformed parameter and transforming it *again* using the area formula. In your example, I'm guessing the *amplitudes* are quite similar between the two peaks, but the *sigma* of the green peak will be HUGE (it's very broad) while the *sigma* of the blue peak comparatively smaller. What you end up with is multiplying the area of the green peak by LOTS and the blue peak by not so much - resulting in the apparently large green peak area compared to the blue.

Now, just in case you don't believe me (totally your perogative!) then I ran through an example where I fit some simulated data with python and derived the area by fitting and integrating, and also compared it with an online calculator by reverse-engineering the fitted gaussian model. You can play with the code here or see a summary here.

Here is my simulated data fitted to an Imfit gaussian model:



Remember this is the Imfit eqn for the gaussian model:

$$f(x;A,\mu,\sigma) = rac{A}{\sigma\sqrt{2\pi}}e^{[-(x-\mu)^2/2\sigma^2]}$$

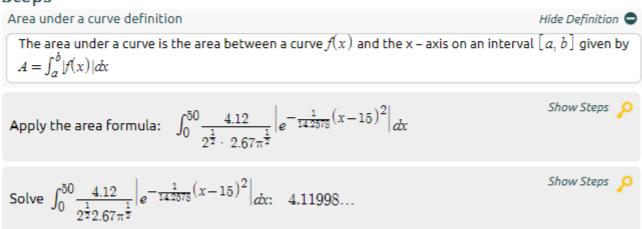
And substituting my fitted parameters into the Imfit equation in the online calculator to determine the area between 0 and 50 via integration:

area 
$$f(x) = \frac{4.12}{2.67\sqrt{2\pi}} e^{\frac{-(x-15.0)^2}{14.2578}}, [0, 50]$$

Notice the A parameter (4.12) is effectively the same as the calculated area (4.119):



## Steps



The area is:

=4.11998...

And just to confirm it is actually calculating the curve we think it is:

