# LEM - Gaussian Hill

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# 1 Landscape Evolution Modeling - Gaussian Hill

#### Bodo Bookhagen

For a more detailed introduction and tutorials, please see the landlab, landlab tutorial, and landlab documentation websites.

### 1.1 Setting up the Model Framework

We will use a Gaussian Hill (as used before) and simulate erosion processes on the Gaussian Hill. We first need to generate a grid and import the Gaussian Hill into the format that can be understand by Landlab. With RasterModelGrid we are generating an empty grid with n\*n pixels, a dem\_width of 5\*100 and a node-spacing of node\_spacing = dem\_width/n - because our Gaussian Hill is centered on 0 and runs from -2.5 to 2.5, but we scale this by a factor of 100 to -25 to 25 (50 width).

You could fill the model grid (mg) with zeros (add\_zeros) via:

```
z = mg.add_zeros('node', 'topographic_elevation')
```

We also create a Gaussian hill (via the function gaussian\_hill\_elevation and save this into the create RasterModelGrid. The Gaussian hill grid has n=111 elements (111x111) and scale the DEM grid to 100m height and 500m width.

mg is now a grid object with 111x111 nodes and 12321 (111x111) nodes and (111x(111-1)x2) links.

```
[2]: mg.number_of_node_columns
```

[2]: 111

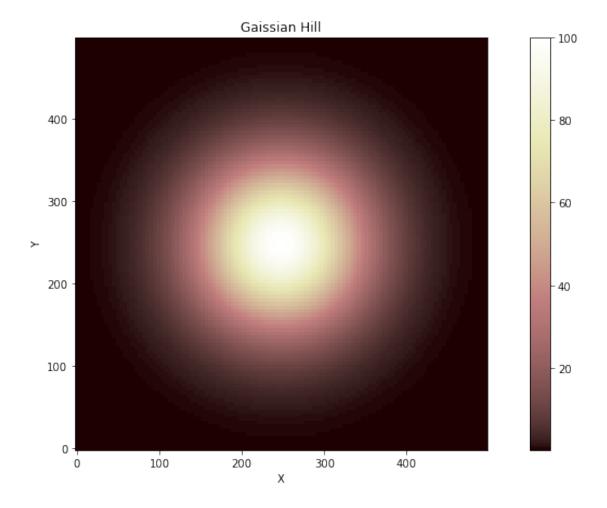
```
[3]: mg.number_of_nodes
```

[3]: 12321

```
[4]: mg.number_of_links
```

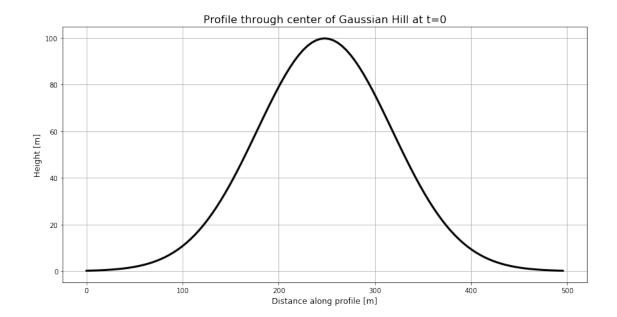
[4]: 24420

Let's plot the grid using the imshow\_grid function from landlab.



In addition, we can plot a profile across the Gaussian Hill:

[6]: Text(0.5, 1.0, 'Profile through center of Gaussian Hill at t=0')



### 1.2 Setting up the Diffusion Modeler

Next, we initiate the boundary conditions for the diffusion modeler. We close all sides and assume a diffusion coefficient of  $0.1m^2/y$ . We run one modeling step that integrates over 1000y.

```
from landlab.components import LinearDiffuser

mg.set_closed_boundaries_at_grid_edges(True, True, True, True)

kappa_ld = 0.1 # m^2/y [L^2/T]

ld = LinearDiffuser(mg, linear_diffusivity=kappa_ld)

dt = 1000 # time step in y

ld.run_one_step(dt)

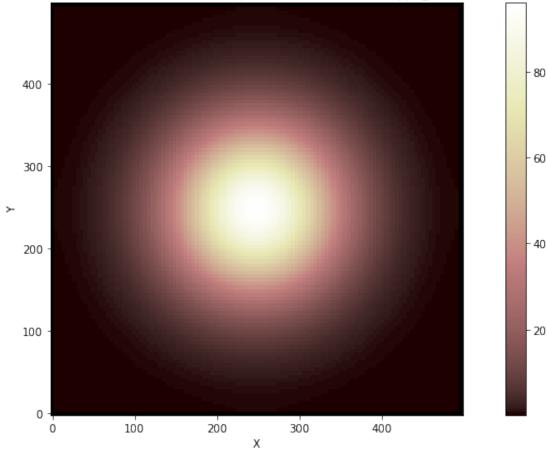
fg3 = pl.figure()

imshow_grid(mg, 'topographic__elevation', plot_name='Gaussian Hill after 1 and_u

→10 time steps with dt=%d and kappa_ld=%f'%(dt, kappa_ld),

allow_colorbar=True)
```



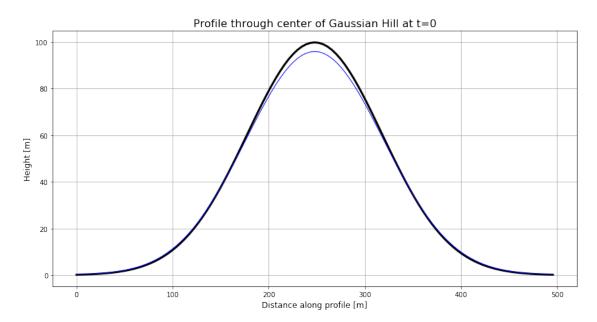


We can save the resulting grid to a separate variable and generate a profile across the diffused Gaussian Hill:

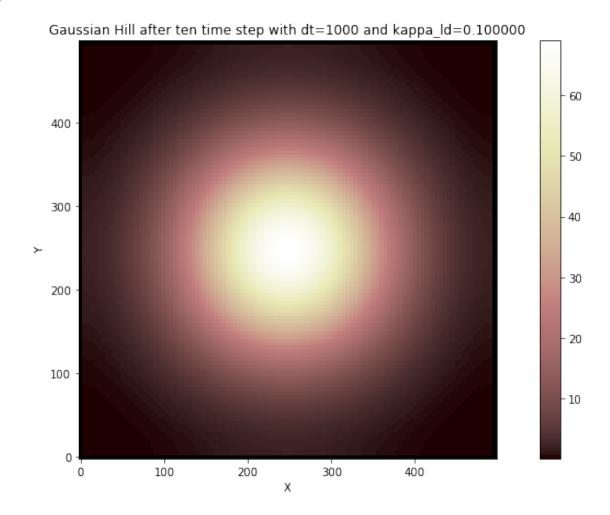
This can be plotted together with the original profile:

```
ax.set_xlabel('Distance along profile [m]', fontsize=12)
ax.set_ylabel('Height [m]', fontsize=12)
ax.set_title('Profile through center of Gaussian Hill at t=0', fontsize=16)
```

### [9]: Text(0.5, 1.0, 'Profile through center of Gaussian Hill at t=0')

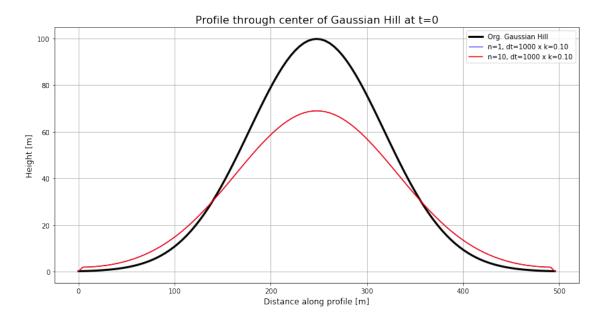


We can use the same setup and run the model through a for loop - here for 10 times - and plot the resulting DEM.



Again, the results after 10 model steps (each  $1000~\mathrm{y}$  - a total of  $10\mathrm{ky}$  model time) can be visualized through profiles:

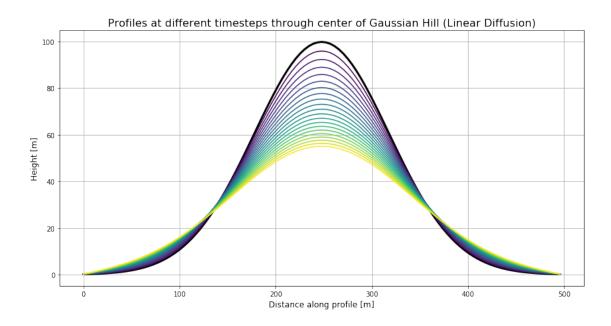
### [11]: <matplotlib.legend.Legend at 0x7f90e83af820>



### 1.3 Plotting each individual Model Results

In the above example, only the final steps are plotted and shown as profile. We can extract the results from each modeling step and can better evaluate the dynamics of the terrain by plotting the results of each time step individually. We do this within the for loop. Let's set up our model again and plot every time step (2000 y steps for 20 times):

```
#re-initiate model domain to start from scratch
x,y,z = gaussian_hill_elevation(n)
z = z*100
mg = RasterModelGrid((n, n), node_spacing)
gh_org = mg.add_field('node', 'topographic__elevation', z, units='meters', __
kappa_1d = 0.05 \# m^2/y [L^2/T]
ld = LinearDiffuser(mg, linear_diffusivity=kappa_ld)
dt = 2000 \# time step in y
time_steps = 20
crosssection_center_dt = np.empty((time_steps, len(crosssection_center_d1)))
crosssection_center_ycoords_dt = np.empty((time_steps,__
→len(crosssection_center_ycoords_d1)))
colors = pl.cm.viridis(np.linspace(0,1,time_steps))
for i in range(time_steps):
   ld.run_one_step(dt)
   gh_ld = mg.node_vector_to_raster(gh_org, flip_vertically=True)
    crosssection_center_dt[i,:] = mg.node_vector_to_raster(gh_ld,__
\rightarrowflip_vertically=True)[:,np.int(np.round(n/2))]
    crosssection_center_ycoords_dt[i,:] = mg.node_vector_to_raster(mg.node_y,_
\rightarrowflip_vertically=True)[:,np.int(np.round(n/2))]
   ax.plot(crosssection_center_ycoords_dt[i,:],
                 crosssection_center_dt[i,:],
                 color=colors[i], label='dt=%d'%(dt))
   print('%d'%(i))
```

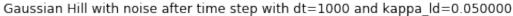


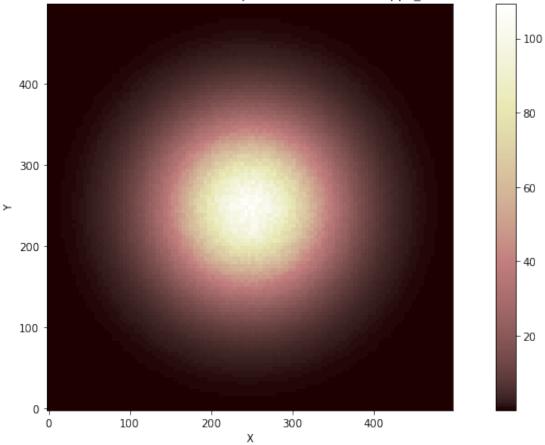
# 1.4 Gaussian Hill with Topographic Noise

We will add some topographic noise to the Gaussian Hill to make a more realistic setup. The Gaussian Hill with noise is plotted below.

```
[13]: x,y,z = gaussian_hill_elevation(n)
z = z*100
z = z + z * np.random.rand(n,n) * 0.1

mg = RasterModelGrid((n, n), node_spacing)
gh_noise = mg.add_field('node', 'topographic_elevation', z, units='meters',u \( \topographic = \topographic
```



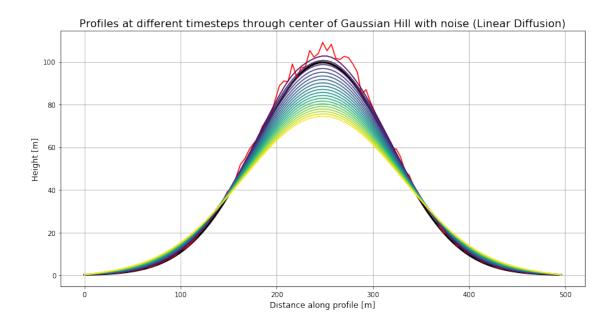


Next, we plot profile for every modeling step (1000 y for 20 time steps).

```
ax.plot(crosssection_center_ycoords_noise_1, crosssection_center_noise_1, 'r', __
→label='GH w/ noise')
crosssection_center_dt = np.empty((time_steps, len(crosssection_center_d1)))
crosssection_center_ycoords_dt = np.empty((time_steps,__
→len(crosssection_center_ycoords_d1)))
colors = pl.cm.viridis(np.linspace(0,1,time_steps))
for i in range(time_steps):
    ld.run_one_step(dt)
    gh_ld = mg.node_vector_to_raster(ghn_ld1, flip_vertically=True)
    crosssection_center_dt[i,:] = mg.node_vector_to_raster(gh_ld,__
\rightarrowflip_vertically=True)[:,np.int(np.round(n/2))]
    crosssection_center_ycoords_dt[i,:] = mg.node_vector_to_raster(mg.node_y,_
\rightarrowflip_vertically=True)[:,np.int(np.round(n/2))]
    ax.plot(crosssection_center_ycoords_dt[i,:], crosssection_center_dt[i,:],__

color=colors[i], label='dt=%d'%(dt))

    print('%d'%(i))
```



Note that the noisy Gaussian Hill surface has been rapidly smoothed.

### 1.5 Adding Fluvial Erosion to a Gaussian Hill through the FastScapeEroder

A description of the FastScapeEroder lists the options. Here, we setup the Gaussian Hill (without noise) and use a K=1e-4 and m=0.5 and n=1 for the stream-power erosion law. We perform the modeling in steps of 5000 y. A standard output of fluvial erosion modeling is the flow accumulation (drainage area) that is shown below. The output of single erosion step is shown in profile form.

```
from landlab.components import FlowAccumulator, FastscapeEroder

x,y,z = gaussian_hill_elevation(n)
z = z*100

mg = RasterModelGrid((n, n), node_spacing)
gh_org = mg.add_field('node', 'topographic_elevation', z, units='meters',u copy=True, clobber=False)

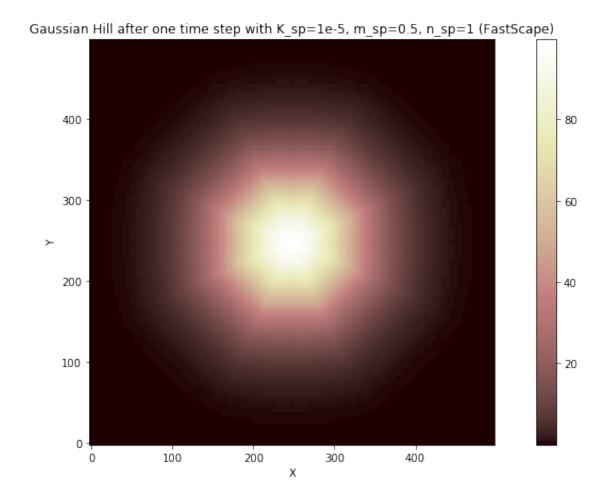
fr = FlowAccumulator(mg, flow_director='D8')
fse = FastscapeEroder(mg, K_sp = 1e-4, m_sp=0.5, n_sp=1.)
fr.run_one_step()
fse.run_one_step(dt=5000.)

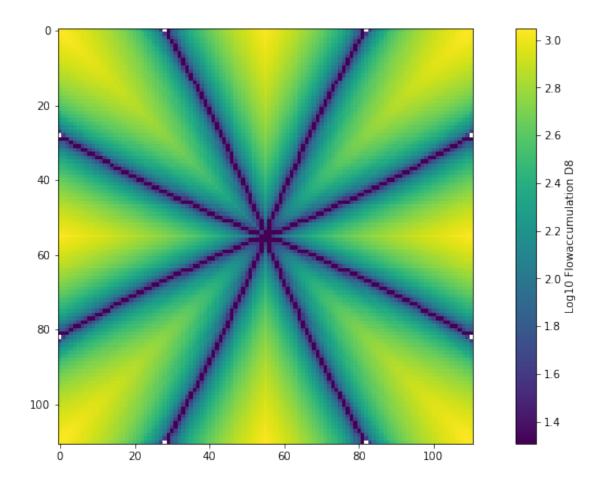
fg7 = pl.figure()
imshow_grid(mg, 'topographic_elevation', plot_name='Gaussian Hill after one_u chime step with K_sp=1e-5, m_sp=0.5, n_sp=1 (FastScape)',
```

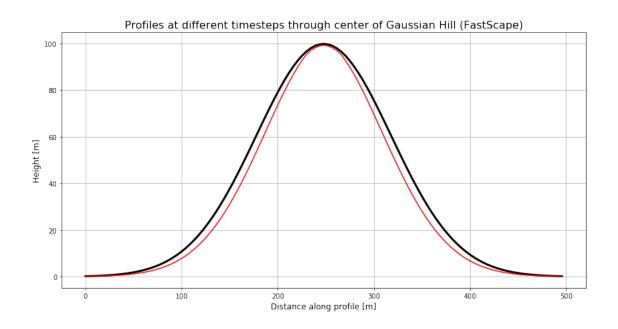
```
allow_colorbar=True)
pl.figure()
pl.imshow(np.log10(np.reshape(fr.node_drainage_area, (n,n))))
cb=pl.colorbar()
cb.set_label('Log10 Flowaccumulation D8')
fg2, ax = pl.subplots(1, 1)
gh_fse = fse.grid.node_vector_to_raster(gh_org, flip_vertically=True)
crosssection_center_fse_1 = mg.node_vector_to_raster(gh_fse,__
→flip_vertically=True)[:,np.int(np.round(n/2))]
crosssection_center_ycoords_fse_1 = mg.node_vector_to_raster(mg.node_y,__
→flip_vertically=True)[:,np.int(np.round(n/2))]
ax.plot(crosssection_center_ycoords, crosssection_center, 'k', linewidth=3,__
→label='Org. Gaussian Hill')
ax.grid()
ax.set_xlabel('Distance along profile [m]', fontsize=12)
ax.set_ylabel('Height [m]', fontsize=12)
ax.set_title('Profiles at different timesteps through center of Gaussian Hill_
ax.plot(crosssection_center_ycoords_fse_1, crosssection_center_fse_1, 'r', __
 →label='noise')
```

```
<ipython-input-15-e90b87772f4b>:19: RuntimeWarning: divide by zero encountered
in log10
   pl.imshow(np.log10(np.reshape(fr.node_drainage_area, (n,n))))
```

[15]: [<matplotlib.lines.Line2D at 0x7f90e6881e80>]





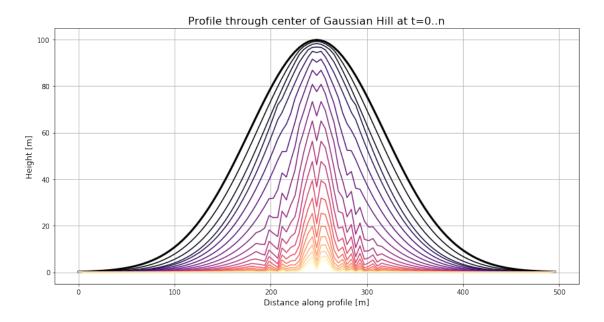


Next, we perform iterative steps and plot a profile across the eroding Gaussian Hill at every step.

```
[16]: x,y,z = gaussian_hill_elevation(n)
      z = z*100
      mg = RasterModelGrid((n, n), node_spacing)
      gh_org = mg.add_field('node', 'topographic_elevation', z, units='meters', __
      →copy=True, clobber=False)
      fr = FlowAccumulator(mg, flow_director='D8')
      fse = FastscapeEroder(mg, K_sp = 1e-4, m_sp=0.5, n_sp=1.)
      fg2, ax2 = pl.subplots(1, 1)
      ax2.plot(crosssection_center_ycoords, crosssection_center, 'k', linewidth=3,__
      →label='Org. Gaussian Hill')
      ax2.grid()
      ax2.set xlabel('Distance along profile [m]', fontsize=12)
      ax2.set_ylabel('Height [m]', fontsize=12)
      ax2.set title('Profile through center of Gaussian Hill at t=0..n', fontsize=16)
      dt=5000.
      time_steps = 20
      crosssection_center_dt = np.empty((time_steps, len(crosssection_center_d1)))
      crosssection_center_ycoords_dt = np.empty((time_steps,__
      →len(crosssection_center_ycoords_d1)))
      colors = pl.cm.magma(np.linspace(0,1,time_steps))
      for i in range(time_steps):
          fr.run one step()
          fse.run_one_step(dt)
          gh_fse = mg.node_vector_to_raster(gh_org, flip_vertically=True)
          crosssection_center_dt[i,:] = mg.node_vector_to_raster(gh_fse,_
       \rightarrowflip_vertically=True)[:,np.int(np.round(n/2))]
          crosssection_center_ycoords_dt[i,:] = mg.node_vector_to_raster(mg.node_y,_
       →flip_vertically=True)[:,np.int(np.round(n/2))]
          ax2.plot(crosssection_center_ycoords_dt[i,:], crosssection_center_dt[i,:],u

color=colors[i], label='i=%d'%(i))

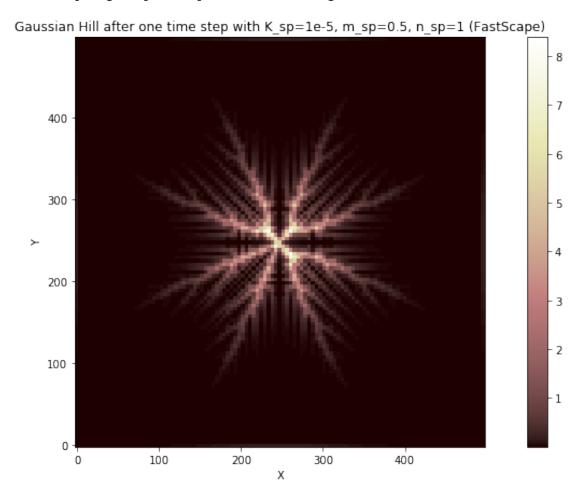
          print('%d'%(i))
```

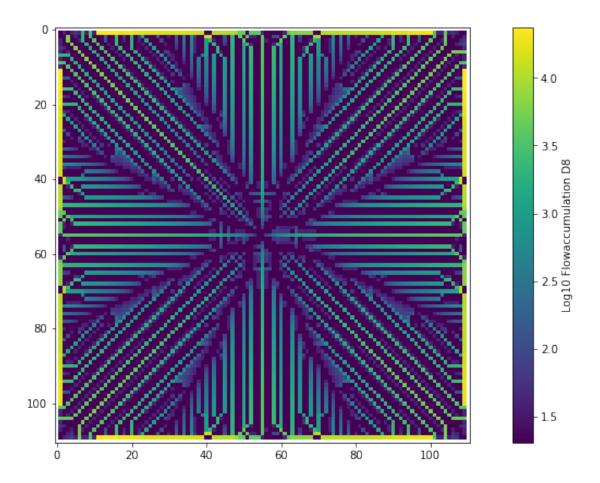


Note the different shape of the resulting landscape and compare it to the profile resoluts from the linear diffusion shown above. The resulting topography and drainage network is plotted below.

<ipython-input-17-7e008c14d7dc>:6: RuntimeWarning: divide by zero encountered in
log10

pl.imshow(np.log10(np.reshape(fr.node\_drainage\_area, (n,n))))





[]: