

Simulation of segmented HPGe detectors in



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Detector Definition

```
In [6]: detector = SolidStateDetector(SSD_examples[:BEGe])
```

BEGe

_____ ExampleSegmentedBEGe _____

---General Properties---

Detector Material: High Purity Germanium
Environment Material: Vacuum
Bulk type: ntype

Out[6]:

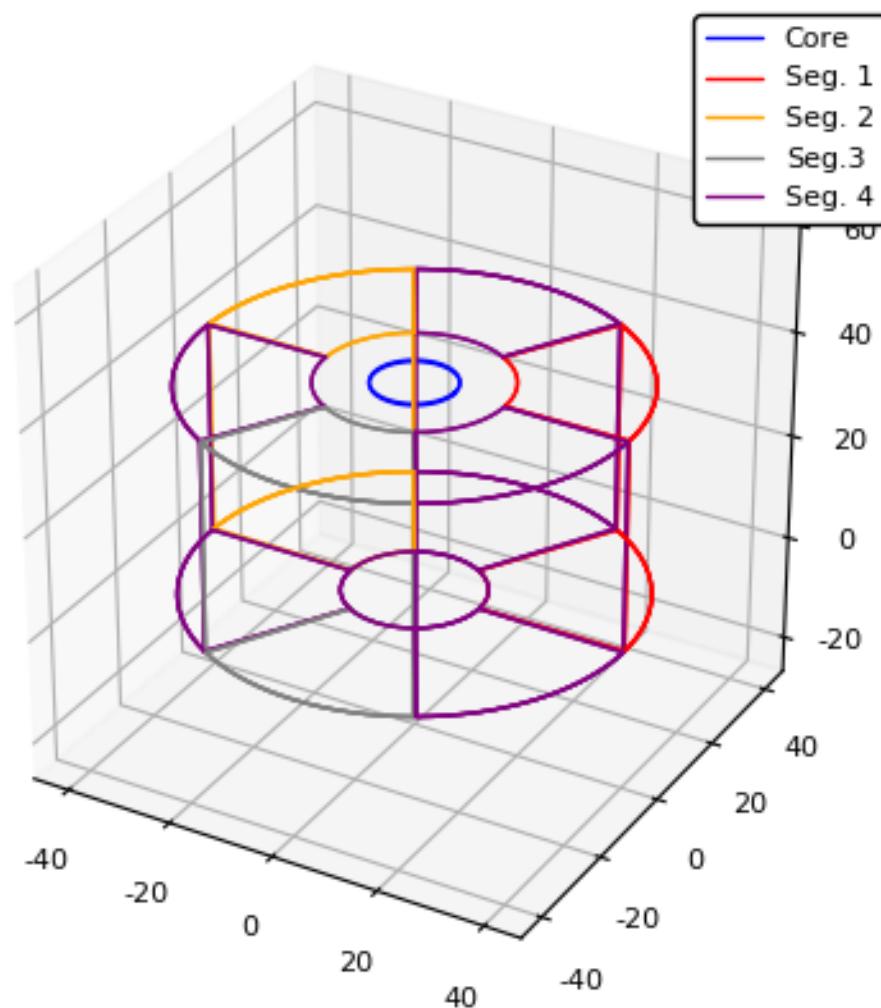
Core Bias Voltage: 4500.0 V
Mantle Bias Voltage: 0.0 V

---Geometry---

Outer Crystal Dimensions:
Crystal length: 40.0 mm
Crystal diameter: 80.0 mm

```
In [7]: plot(detector)
```

Out[7]:



Electric Potential Calculation

To simulate charge drift, we need the E-field. For the E-field, we need E_{pot} . Simple physics problem:

$$\left(\frac{\partial}{\partial x^2} + \frac{\partial}{\partial y^2} + \frac{\partial}{\partial z^2} \right) E_{pot} = \frac{\rho(x, y, z)}{\epsilon_0}$$

Relaxation algorithm (simplified):

- Discretize problem by creating a grid of spacial points
- For each point:
 - Determine potential divergence numerically
 - Compare with $\rho(x, y, z)/\epsilon_0$
 - Apply correction to potential value of current point
 - Repeat until converged
- Challenge: Achieve numerical stability and fast convergence

Choosing a smart grid and strategy

- Red-black grid: Update only odd or even points in each cycle
- Use over-relaxation for faster convergence

In [9]: rb_plot

Out [9]:

3D Electrical Potential Calculation

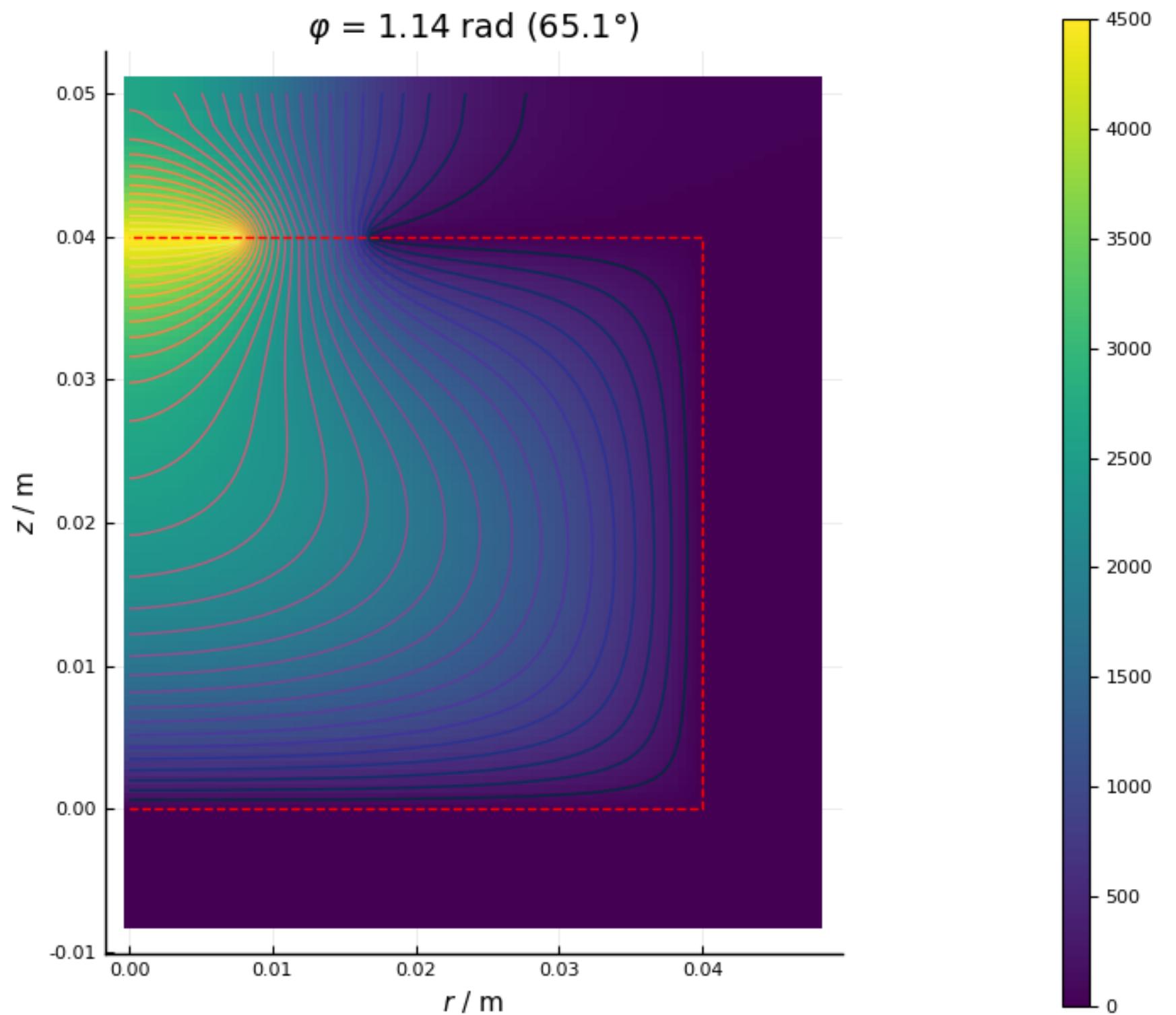
- Full 3D potential takes longer than in 2D
- But: Can now store and load potentials (HDF5)

```
In [10]: E_pot, point_types = if !isfile("detector-potentials-segbege.h5")
          println("Calculating electrical potential, this may take a while ...")
          E_pot, point_types = calculate_electric_potential(detector, convergence_limit=1e-
5,max_refinements=2)
else
    println("Reading electrical potential from file.")
    HDF5.h5open("detector-potentials-segbege.h5", "r") do input
        CylindricalGrid(readdata(input, "detpot/E_pot")),
        PointTypes(readdata(input, "detpot/pointflags"))
    end
end;
```

Reading electrical potential from file.

```
In [98]: plot(E_pot, ϕ=1.12, contours_equal_potential=true, levels=30, size = (800, 600))  
plot!(detector, :plane, ϕ=0)
```

Out[98]:



Simulating Undepleted Detectors

```
In [80]: detector_undep = deepcopy(detector)
detector_undep.segment_bias_voltages[1] = 1500;
```

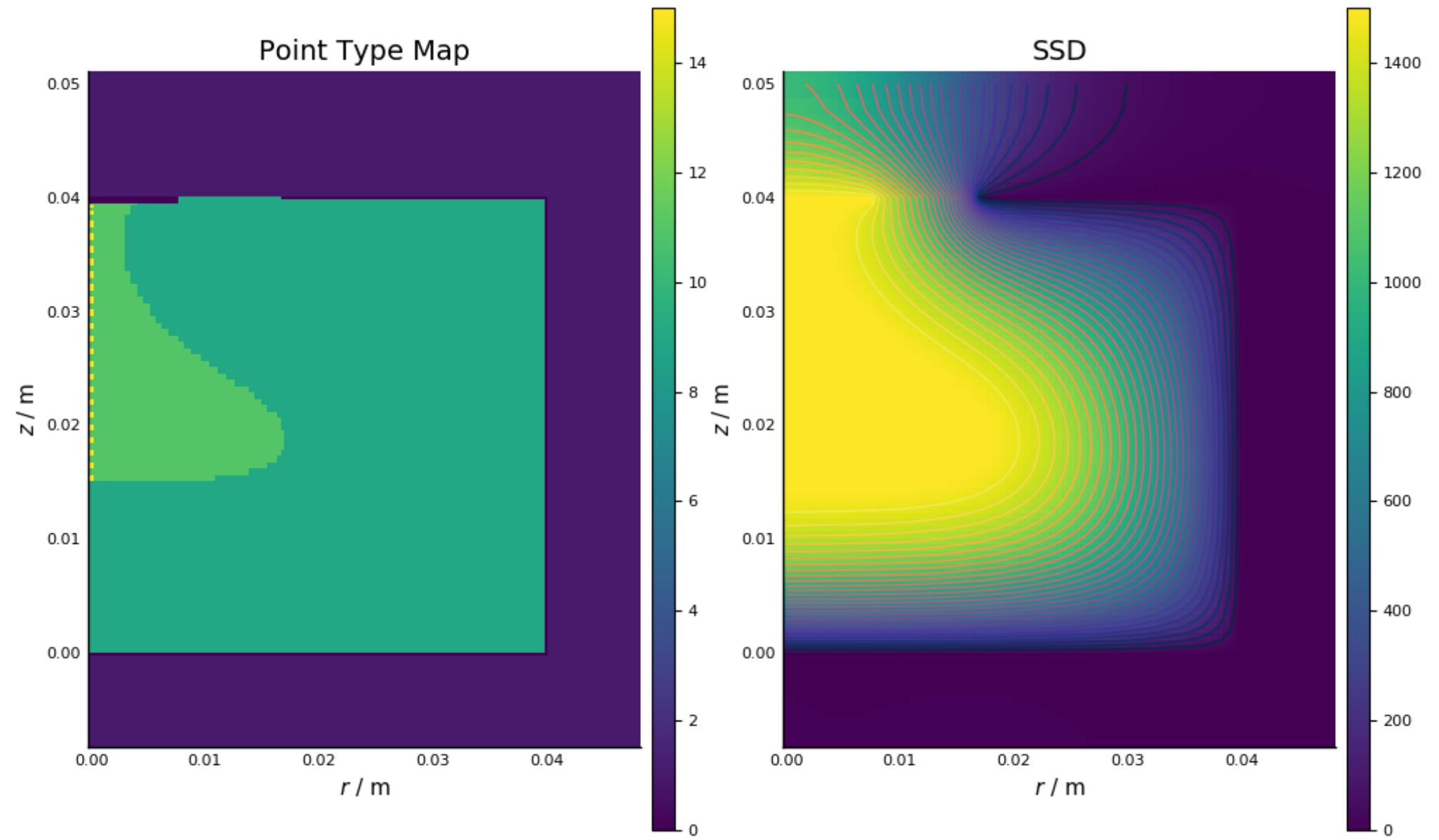
```
In [97]: E_pot_undep, point_types_undep = calculate_electric_potential(
    detector_undep, depletion_handling=true,
    convergence_limit=1e-5, max_refinements=2
);
```

```
Electric Potential Calculation
Bulk type: ntype
Bias voltage: 1500.0 V
φ symmetry: cyclic = 120.0° -> just calculating 1/3 of the detector in φ.
Precision: Float32
Convergence limit: 1.0e-5 => 0.015 V
Threads: 4
Refine? -> true
Refinement parameters:
    maximum number of refinements: 2
    minimum grid spacing:
        r: 0.0001 m
        φ: 0.01 rad
        z: 0.0001 m
    Refinement limits:
        r: 0.0001 -> 0.15 V
        φ: 0.0001 -> 0.15 V
        z: 0.0001 -> 0.15 V
```

```
Convergence: (thresh = 0.015, value = 61.712)
```

```
In [101]: p_SSD = plot(E_pot_undep, title="SSD", contours_equal_potential=true, φ=deg2rad(40),  
levels=30)  
p_pts = plot(point_types_undep, title="Point Type Map", φ=deg2rad(40))  
plot(p_pts, p_SSD, layout=@layout [a b]), size=(1000,600))
```

Out[101]:



Next Step: Calculate Electric Field from Potential

```
In [17]: E_field = SolidStateDetectors.get_electric_field_from_potential(E_pot);
```

```
In [85]: plot(E_field, detector, E_pot, size = (800, 600));  
plot!(detector, :plane)
```

Out[85]:

Calculate Drift Fields

- Precalculation of drift vector fields increases performance of charge drift

```
In [19]: drift_model = ADLChargeDriftModel(); # drift_model_vacuum = VacuumChargeDriftModel()
()

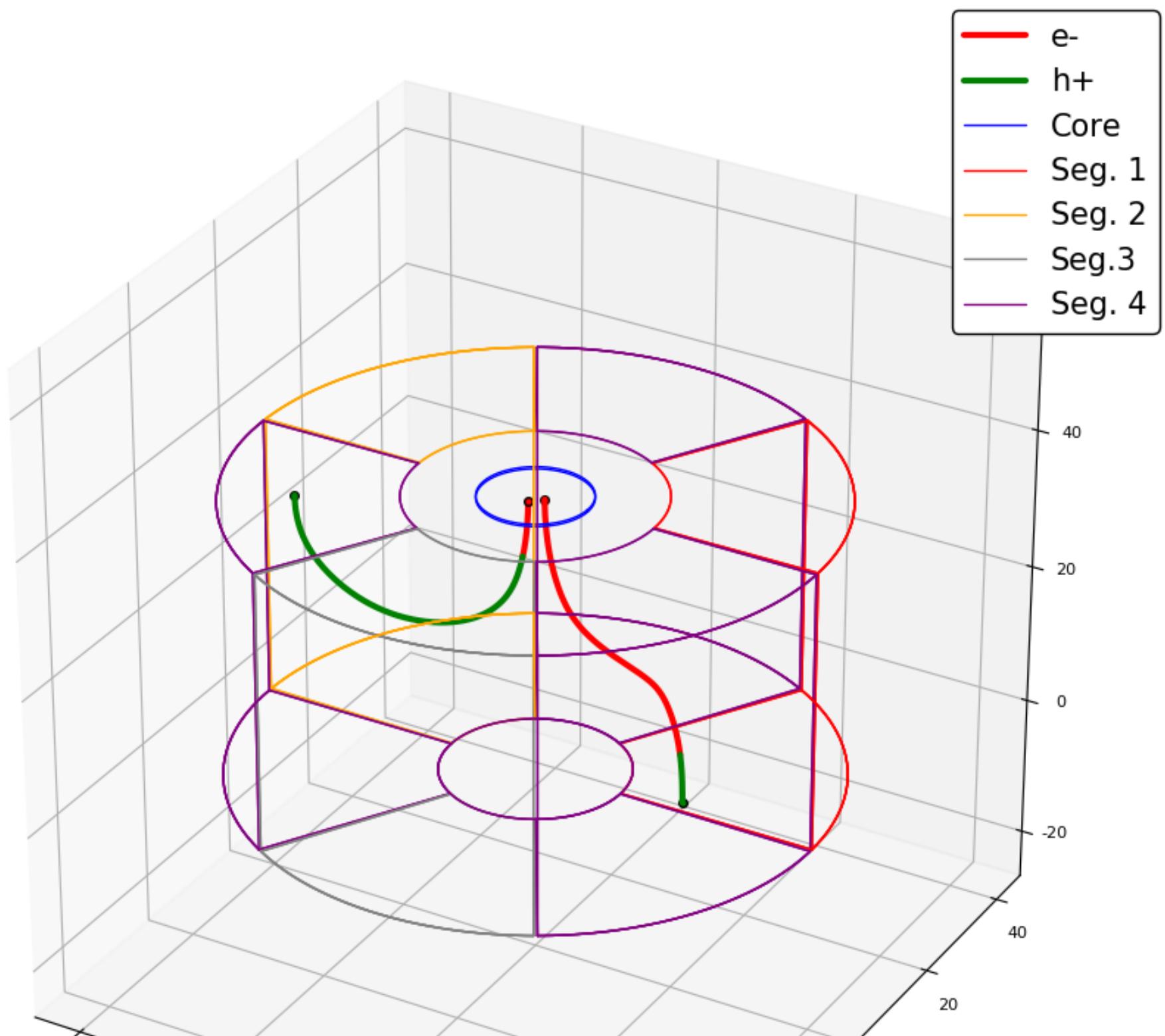
In [20]: electron_drift_field = get_electron_drift_field(E_field, drift_model);
hole_drift_field = get_hole_drift_field(E_field, drift_model);

In [21]: # Interpolate
electron_drift_field_interpolated = SolidStateDetectors.get_interpolated_drift_field
(electron_drift_field, E_pot);
hole_drift_field_interpolated = SolidStateDetectors.get_interpolated_drift_field(hol
e_drift_field, E_pot);

In [23]: T = Float32
pos = [SVector{3,T}(-0.0007,- 0.002,0.032),SVector{3,T}(0.02, 0.002,0.007)]
drift_paths=SolidStateDetectors.drift_charges(detector, pos, electron_drift_field_in
terpolated, hole_drift_field_interpolated);
```

```
In [24]: plot(drift_paths, scaling=1e3)
plot!(detector, size = (900, 900))
```

Out [24]:



Weighting Potential Calculation

- Detector has a point contact and 4 segments - takes a few minutes
- But: Can now store to and load from HDF5:

```
In [25]: wps = if !isfile("detector-potentials-segbege.h5")
    println("Calculating weighting potentials, this may take a while ...")
    [SolidStateDetectors.calculate_weighting_potential(detector, i, verbose=false, ma
x_refinements=4, max_n_iterations=500, convergence_limit=1e-5) for i in 1:5]
else
    println("Reading weighting potentials from file.")
    wps = HDF5.h5open("detector-potentials-segbege.h5", "r") do input
        [CylindricalGrid(readdata(input, "detpot/W_pot/$i")) for i in 1:detector.n_t
otal_contacts]
    end
end;
```

Reading weighting potentials from file.

Software Documentation

- Still needs work, but:

```
In [28]: ?calculate_weighting_potential
```

```
search: calculate_weighting_potential
```

```
Out[28]: calculate_weighting_potential(det::SolidStateDetector, channels::Array{Int, 1};  
    <keyword arguments>) ::Grid
```

Compute the weighting potential of the `channels` of the given Detector `det` on an adaptive grid through successive over relaxation. `channels` is a list of the channels which are fixed to 1. All other channels are fixed to 0.

There are several `<keyword arguments>` which can be used to tune the computation:

Keywords

- `coordinates::Symbol`: the kind of the coordinate system of the grid. Right now only `:cylindrical` is possible.
- `convergence_limit::Real`: `convergence_limit` times the bias voltage sets the convergence limit of the relaxation. The convergence value is the absolute maximum difference of the potential between two iterations of all

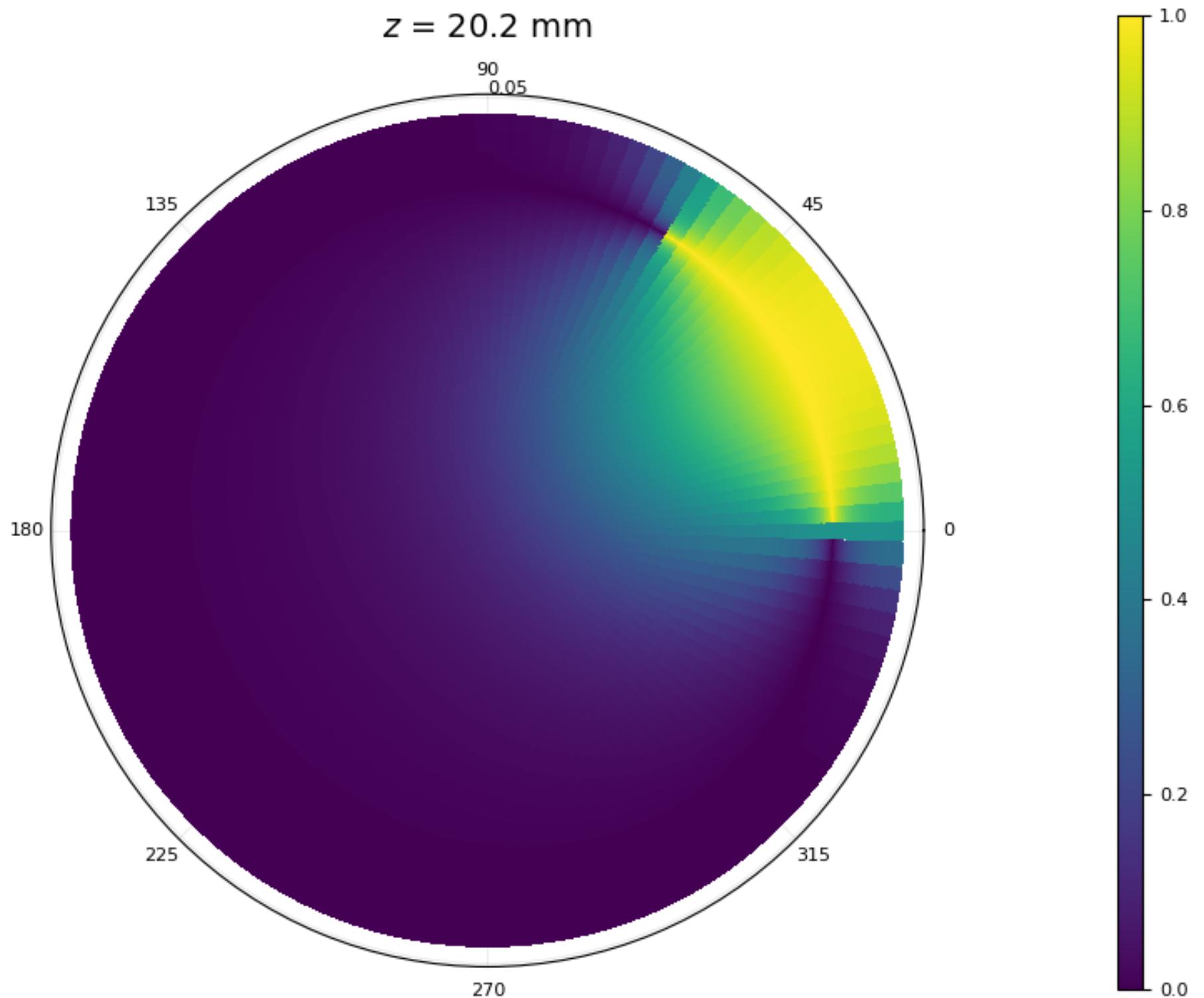
Segment Weights

```
In [89]: seg_wpots_plot
```

```
Out[89]:
```

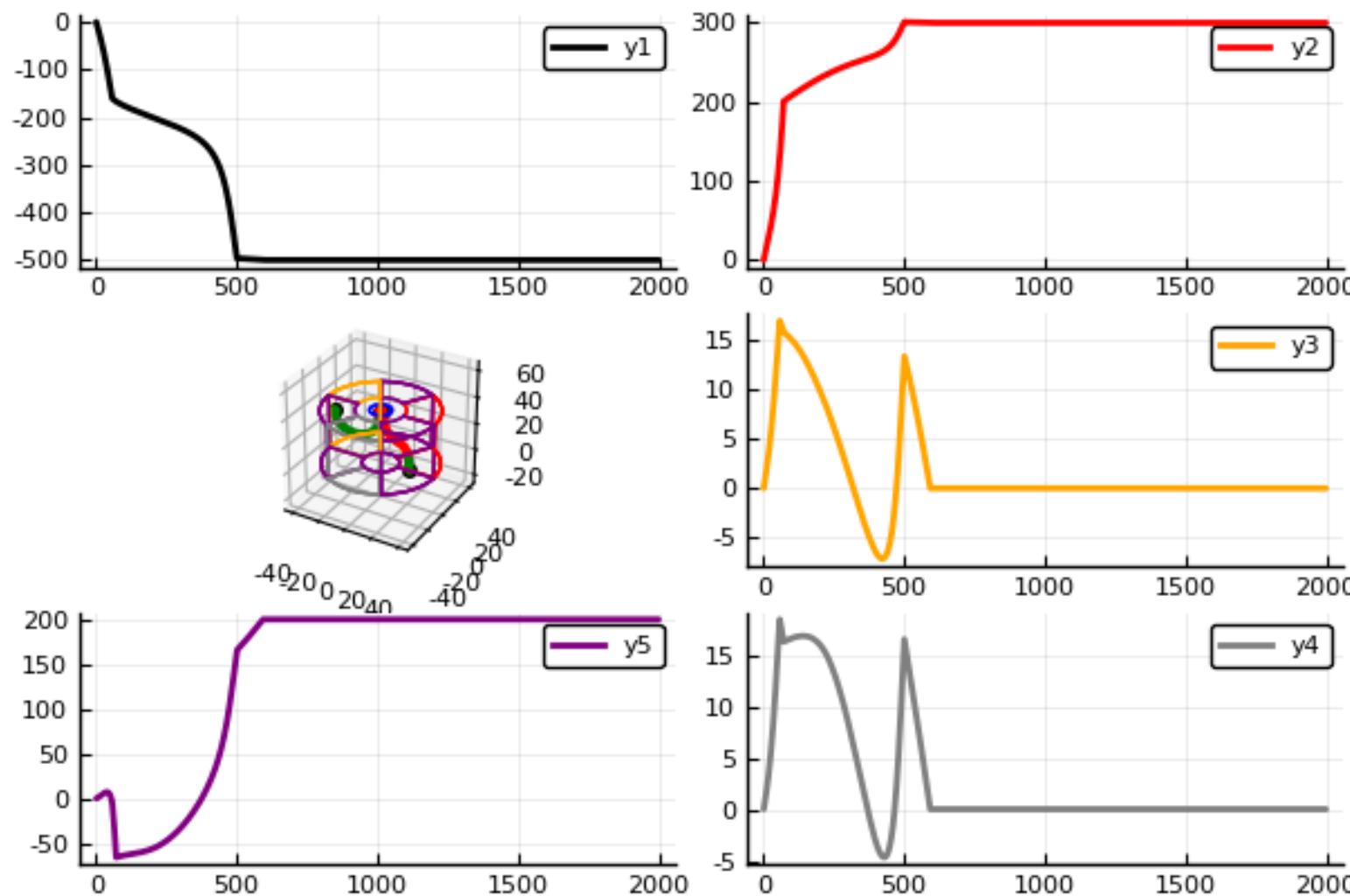
```
In [90]: plot(wps[2], z = 0.02, size = (800, 600))
```

Out[90]:



In [75]: seg_pulses_plot

Out [75]:



Let's load some MC data (from HDF5)

```
In [36]: mctruth_filename = "seg-bege-mctruth.h5"
mc_events = @time HDF5.h5open(mctruth_filename, "r") do input
    readdata(input, "mctruth")
end
println("$length(flatview(mc_events.edep))) original hits")

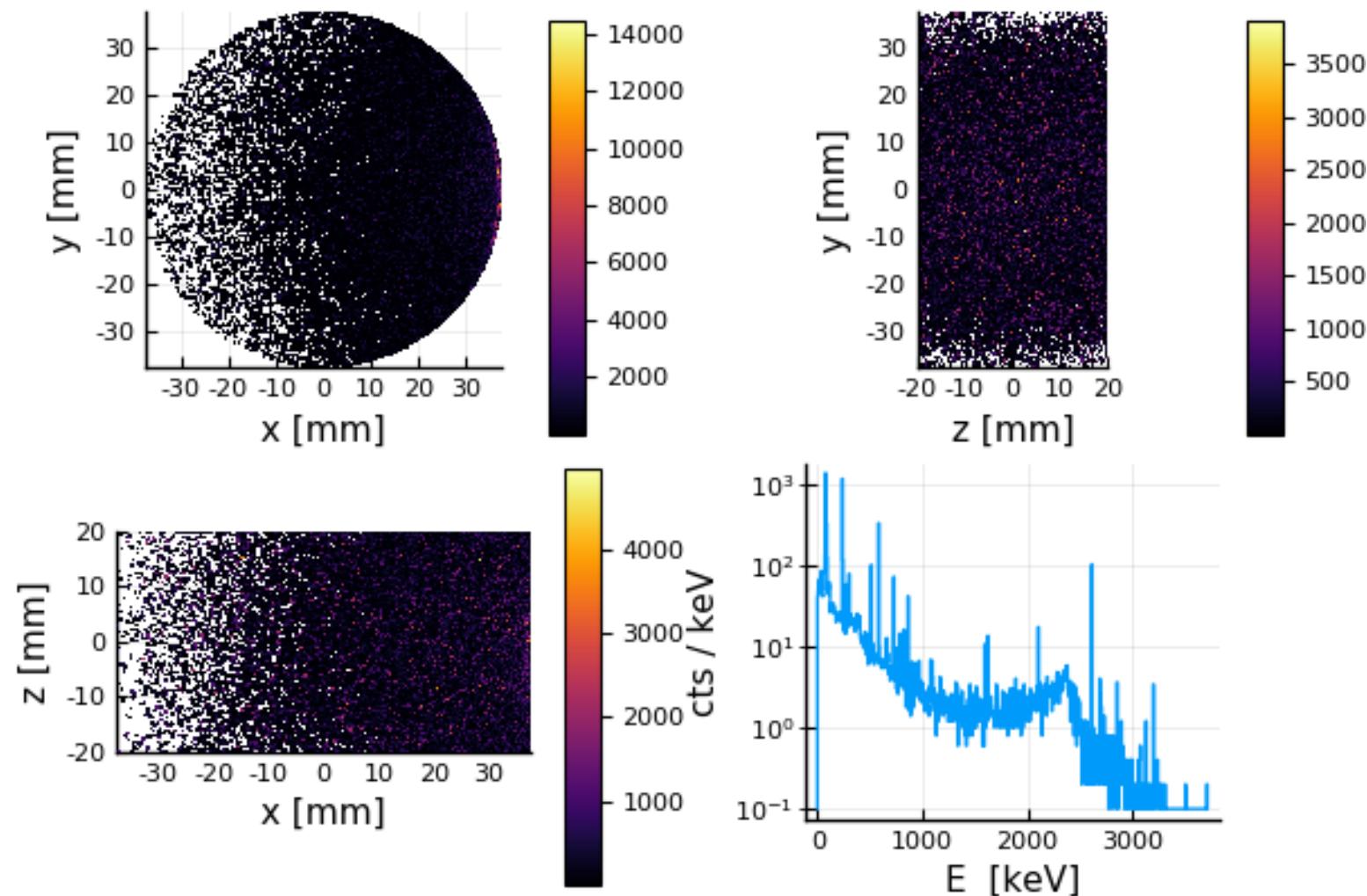
mc_events_clustered = @time cluster_detector_hits(mc_events, 0.2u"mm")
println("$length(flatview(mc_events_clustered.edep))) clustered hits")
```

```
1.226765 seconds (3.08 M allocations: 221.108 MiB, 7.05% gc time)
517456 original hits
4.050771 seconds (16.96 M allocations: 1.114 GiB, 16.82% gc time)
81097 clustered hits
```

Hit Distribution

```
In [37]: plot(mc_events_clustered)
```

Out [37]:

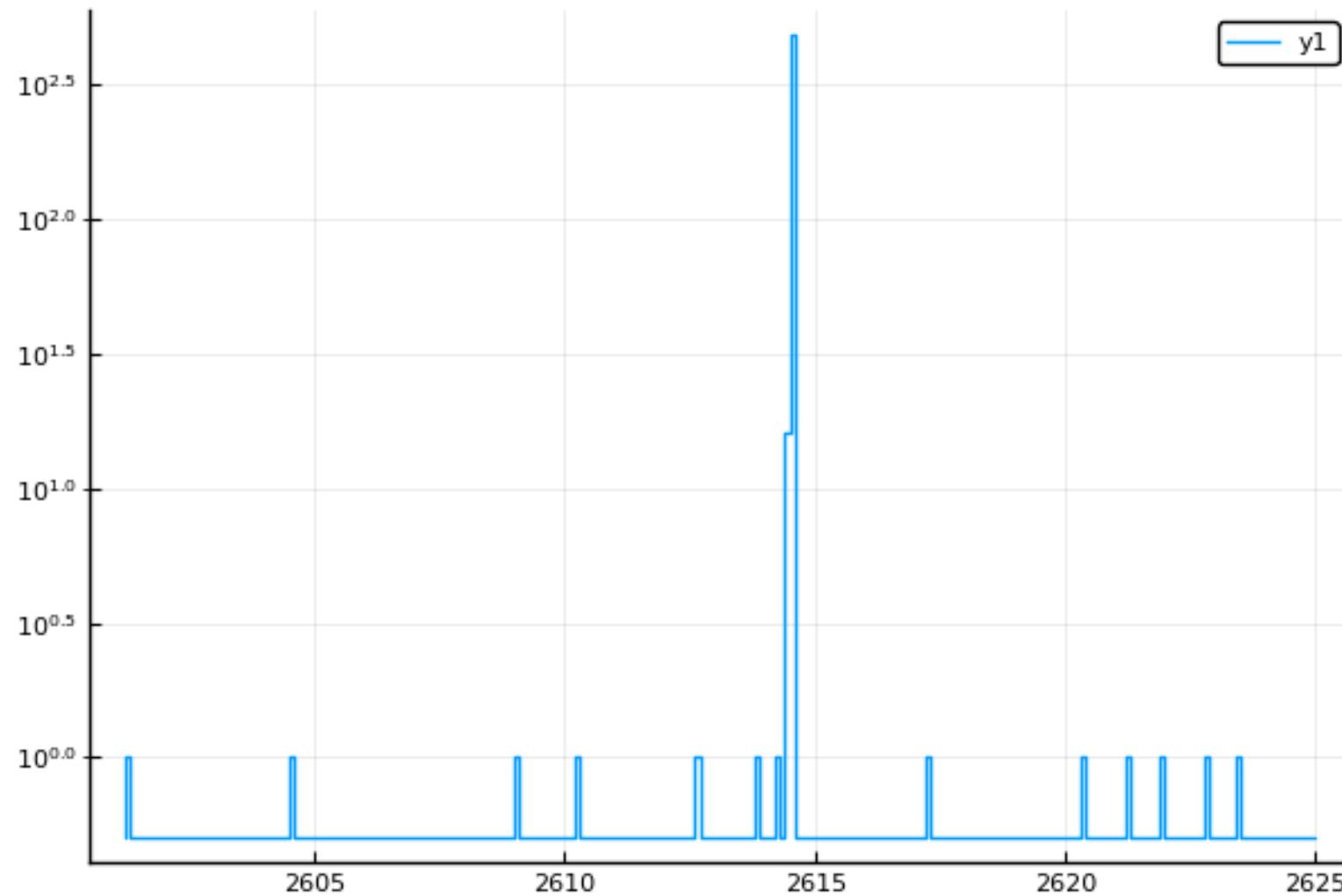


Raw MC Spectrum

- Not very realistic yet ...

```
In [38]: stephist(ustrip.(sum.(mc_events_clustered.edep)), bins = 2600:0.1:2625, yscale = :log10)
```

Out [38]:

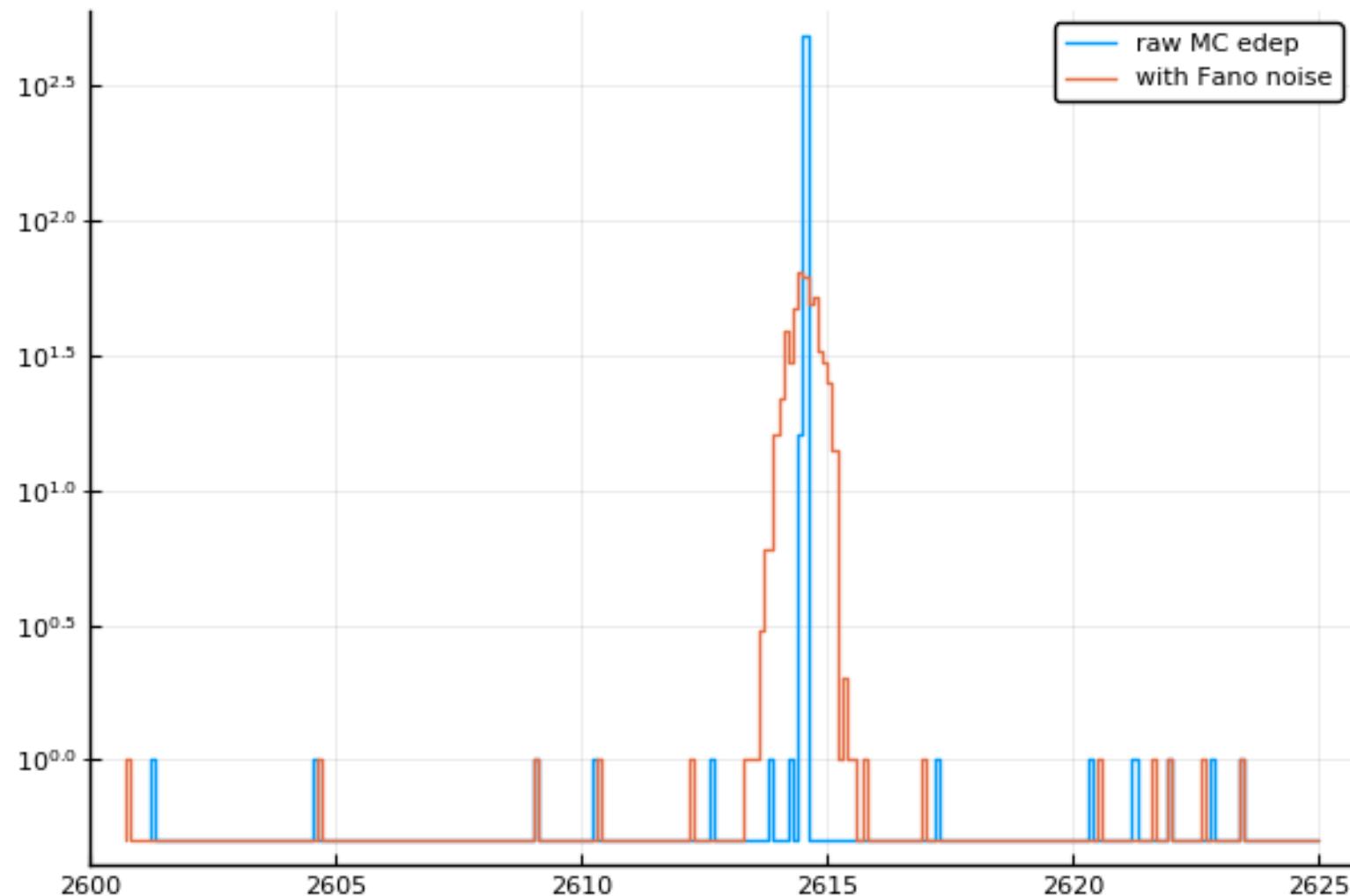


More Realism: Add Fano noise

In [39]:

```
det_material = detector.material_detector
mc_events_fnoise = add_fano_noise(mc_events_clustered, det_material.E_ionisation, de-
t_material.f_fano)
stephist(ustrip.(sum.(mc_events_clustered.edep)), bins = 2600:0.1:2625, label = "raw
MC edep", yscale = :log10)
stephist!(ustrip.(sum.(mc_events_fnoise.edep)), bins = 2600:0.1:2625, label = "with
Fano noise", yscale = :log10)
```

Out [39]:



Filter out hits outside of detector

```
In [50]: filtered_events = mc_events_fnoise[findall(pts -> all(p -> p in detector, pts), mc_events_fnoise.pos)]
```

Out [50]: Table with 5 columns and 16856 rows:

	evtno	detno	thit	edep	...
1	3	[1]	Quantity{Float64,T,...	Quantity{Float64,L^...	...
2	5	[1, 1]	Quantity{Float64,T,...	Quantity{Float64,L^...	...
3	7	[1, 1]	Quantity{Float64,T,...	Quantity{Float64,L^...	...
4	10	[1, 1, 1, 1]	Quantity{Float64,T,...	Quantity{Float64,L^...	...
5	61	[1]	Quantity{Float64,T,...	Quantity{Float64,L^...	...
6	69	[1, 1]	Quantity{Float64,T,...	Quantity{Float64,L^...	...
7	81	[1, 1]	Quantity{Float64,T,...	Quantity{Float64,L^...	...
8	95	[1]	Quantity{Float64,T,...	Quantity{Float64,L^...	...
9	119	[1]	Quantity{Float64,T,...	Quantity{Float64,L^...	...
10	135	[1]	Quantity{Float64,T,...	Quantity{Float64,L^...	...
11	143	[1, 1, 1]	Quantity{Float64,T,...	Quantity{Float64,L^...	...
12	157	[1]	Quantity{Float64,T,...	Quantity{Float64,L^...	...
13	159	[1, 1]	Quantity{Float64,T,...	Quantity{Float64,L^...	...
14	166	[1]	Quantity{Float64,T,...	Quantity{Float64,L^...	...
15	194	[1]	Quantity{Float64,T,...	Quantity{Float64,L^...	...
16	219	[1]	Quantity{Float64,T,...	Quantity{Float64,L^...	...
17	222	[1, 1, 1]	Quantity{Float64,T,...	Quantity{Float64,L^...	...
18	233	[1]	Quantity{Float64,T,...	Quantity{Float64,L^...	...
19	234	[1]	Quantity{Float64,T,...	Quantity{Float64,L^...	...
20	243	[1, 1]	Quantity{Float64,T,...	Quantity{Float64,L^...	...
21	244	[1]	Quantity{Float64,T,...	Quantity{Float64,L^...	...
22	302	[1, 1]	Quantity{Float64,T,...	Quantity{Float64,L^...	...
23	316	[1]	Quantity{Float64,T,...	Quantity{Float64,L^...	...
:	:	:	:	:	.

Generate Pulses for all MC Events

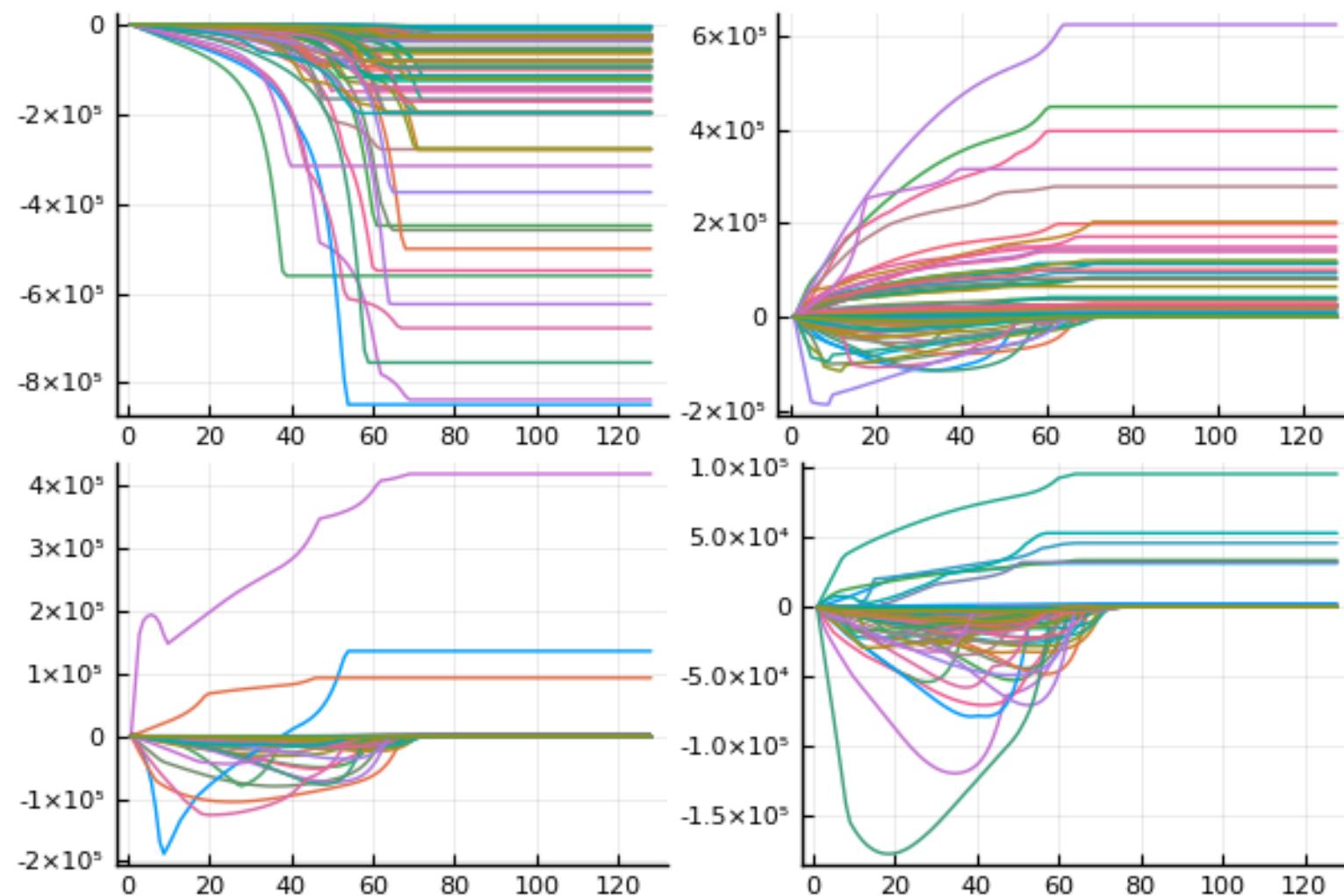
```
In [52]: delta_t = 10u"ns"
contact_charge_signals, drift_times = @time generate_charge_signals(
    detector, electron_drift_field_interpolated, hole_drift_field_interpolated, wps_
    interp,
    filtered_events,
    128, delta_t
);
```

8.866608 seconds (15.11 M allocations: 328.088 MiB, 2.63% gc time)

Let's take a look at the pulses

```
In [53]: plot([plot(contact_charge_signals[i][1:100], legend = false) for i in 1:4]...)
```

Out [53]:

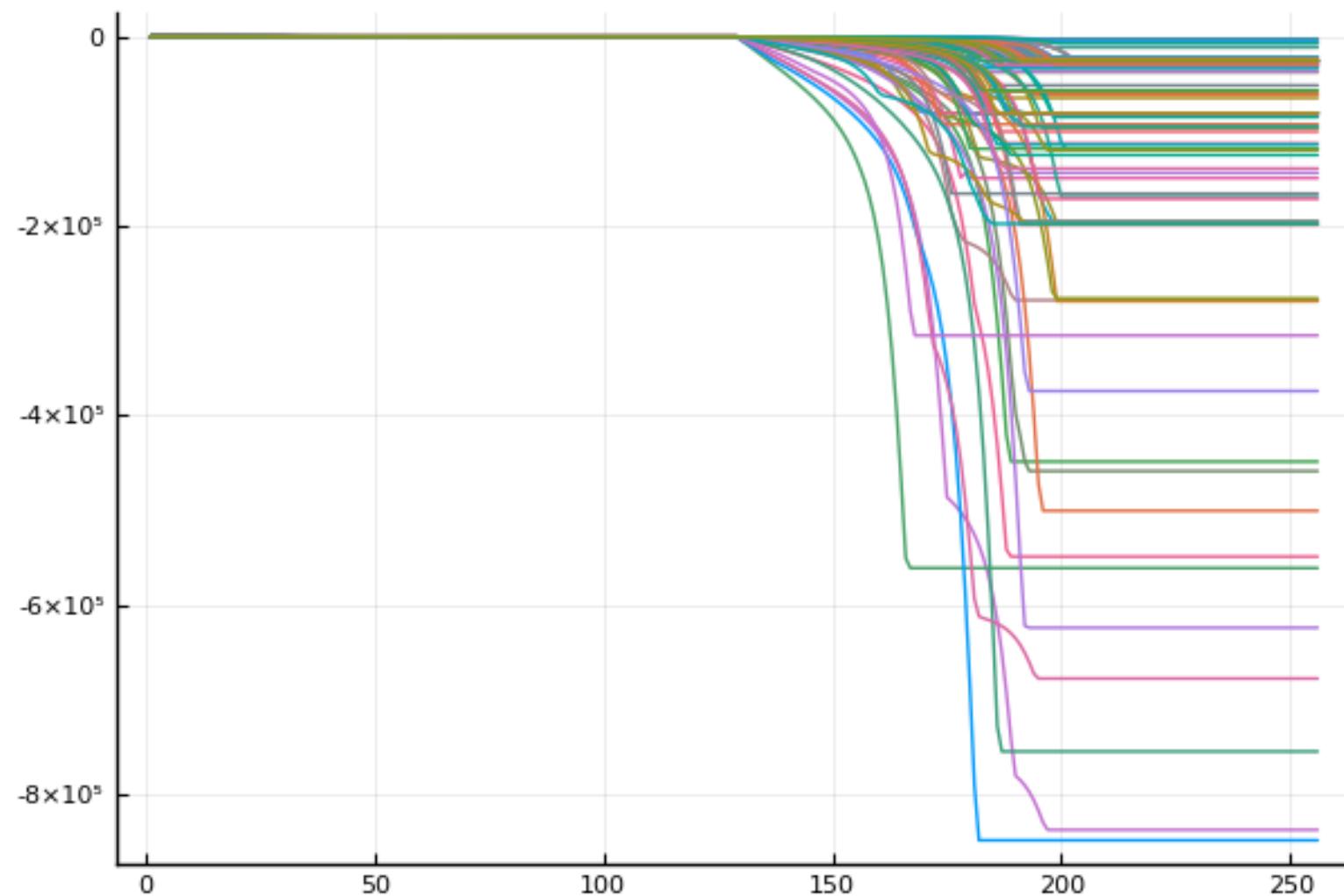


Adding a Pre-Pulse Baseline

In [54]:

```
core_signals = contact_charge_signals[1]
signals_with_baseline = nestedview(vcat(fill!(similar(flatview(core_signals)), 0), f
latview(core_signals)))
plot(signals_with_baseline[1:100], legend = false)
```

Out [54]:



Electronics are Noisy

```
In [95]: noise = rand!(Normal(0,5e3), similar(flatview(signals_with_baseline)))
noisy_signals = nestedview(flatview(signals_with_baseline) .+ noise)
plot(noisy_signals[1:100], legend = false)
```

Out [95]:

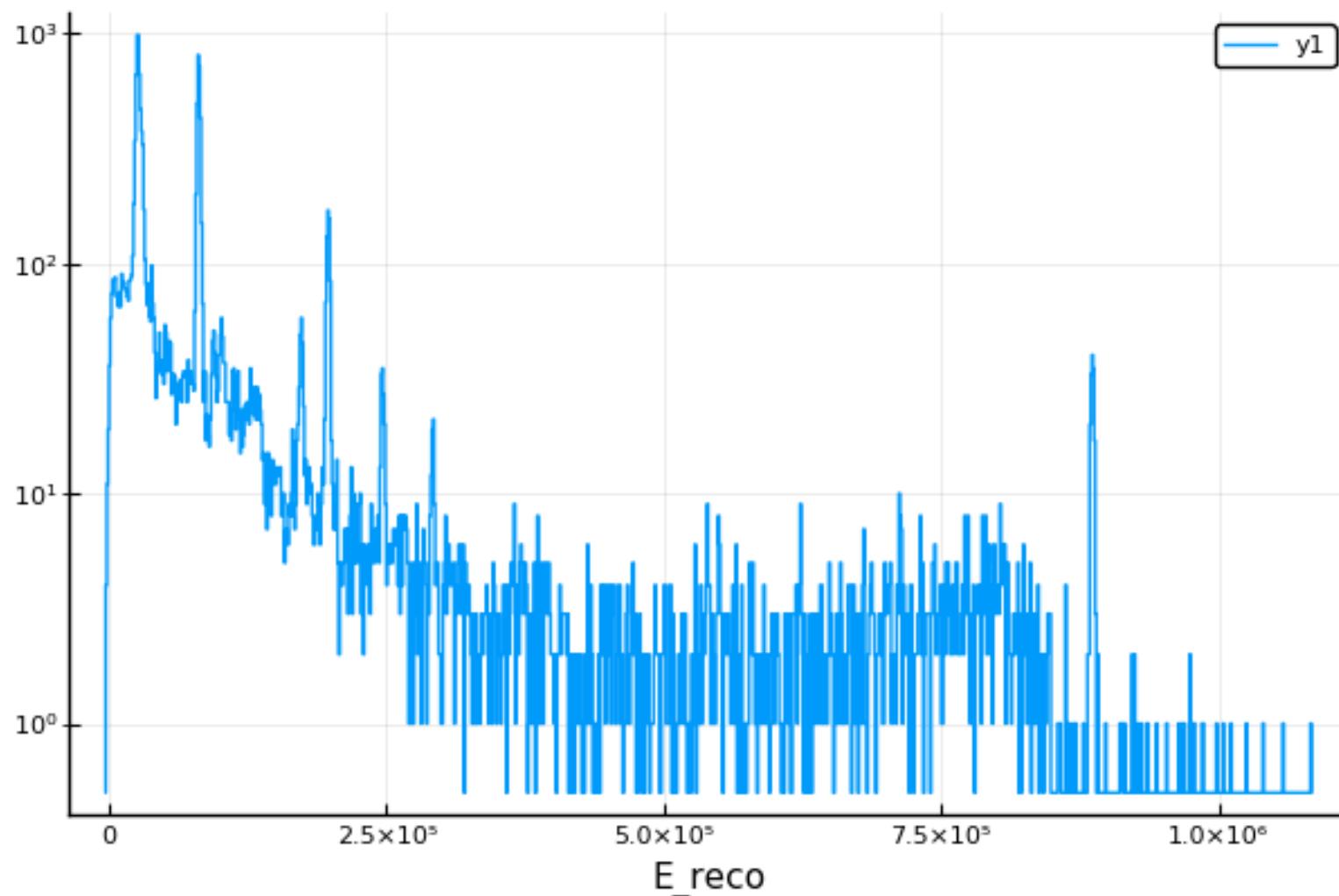


(Simple) Energy Reconstruction on Simulated Pulses

In [96]:

```
pre_pulse_mean = vec(mean(-flatview(noisy_signals)[1:30, :], dims = 1))
post_pulse_mean = vec(mean(-flatview(noisy_signals)[230:255, :], dims = 1))
E_reco = post_pulse_mean .- pre_pulse_mean
stephist(E_reco, nbins = 1000,yscale = :log10, xlabel = "E_reco")
```

Out [96]:



Simulation Verification and Detector Bulk Behaviour

- Simulation looks nice, but is it real?
- Surface scans provide part of the answer (drift times, etc.)
- But how to "look inside" of the detector?
- Common technique: 90° Compton-scanning (Agata, Gretina, etc.), but slow and bulky setup
- What if we were not restricted to 90°? Would increase efficiency a lot!

New MPP Compton Scanner Setup

MPP Compton Scanner, First Data

