

# multimessenger/supernova\*

Tool for calculating CEvNS interactions

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**SNEWS Collaboration Meeting**  
**05.08.2022 Thursday**

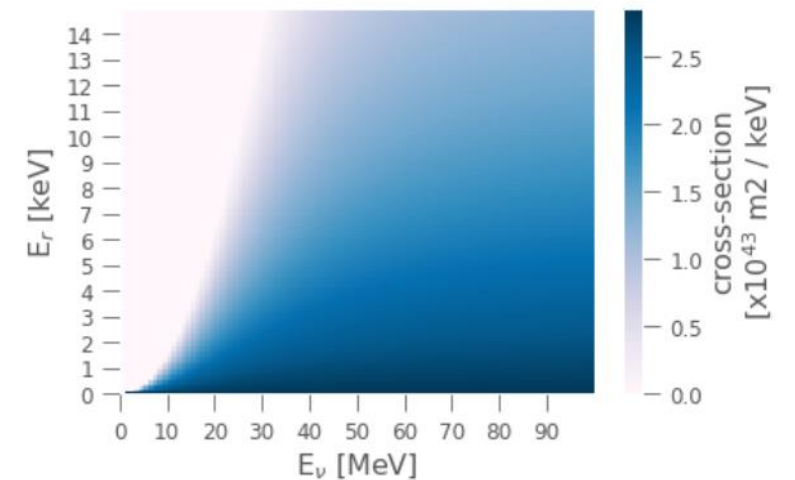
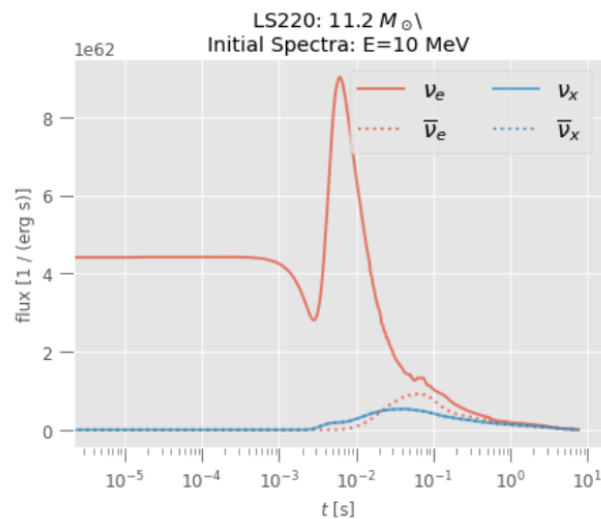
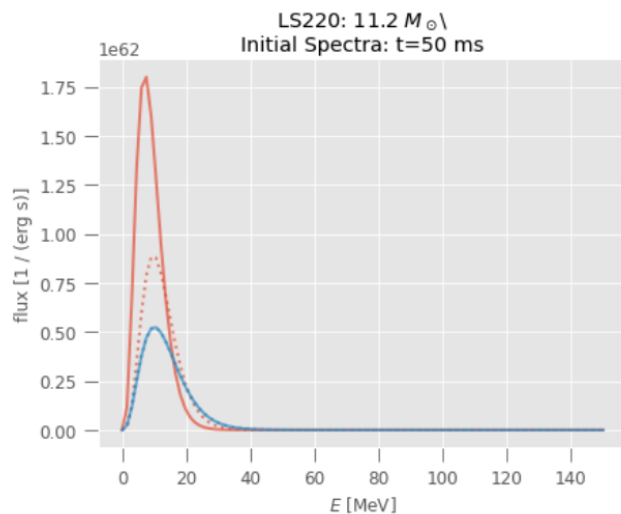
\* cause we are running out of abbreviations

# CEvNS interaction Rates

$$\frac{d^2 R}{dE_R dt_{pb}} = \sum_{\nu_\beta} N_{Xe} \int_{E_{min}^\nu} dE_\nu f_\nu(E_\nu, t, d) \frac{d\sigma}{dE_R}(E_\nu, E_R)$$

fluxes

Cross-sections



# CEvNS interaction Rates

```
A = sn.Models(model_name='Fornax_2021')
```

> Available files for this model, please select an index

```
[0] lum_spec_12M_r10000_dat.h5
[1] lum_spec_13M_r10000_dat.h5
[2] lum_spec_14M_r10000_dat.h5
[3] lum_spec_15M_r10000_dat.h5
[4] lum_spec_16M_r10000_dat.h5
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[13] lum_spec_26.99M_r10000_dat.h5
[14] lum_spec_26M_r10000_dat.h5
```

- Object oriented
- Flexible (change the input composite)
- Scalable by the distance
  
- Computes rates per time,

Fluxes can be fetched from any snewpy model

$$\frac{d^2 R}{dE_R dt_{pb}} = \sum_{\nu\beta} N_{Xe} \int_{E_{min}^{\nu}} dE_{\nu} f_{\nu}(E_{\nu}, t, d) \frac{d\sigma}{dE_R}(E_{\nu}, E_R)$$

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```

- Object oriented
- Flexible (change the input composite)
- Scalable by the distance
  
- Computes rates per time, rates per recoil energy

Fluxes can be fetched from any snewpy model

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```

- Object oriented
- Flexible (change the input composite)
- Scalable by the distance
  
- Computes rates per time, rates per recoil energy per neutrino flavor

Fluxes can be fetched from any snwpy model

$$\frac{d^2 R}{dE_R dt_{pb}} = \sum_{\nu_\beta} N_{Xe} \int_{E_{min}^\nu} dE_\nu f_\nu(E_\nu, t, d) \frac{d\sigma}{dE_R}(E_\nu, E_R)$$

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```

- Object oriented
- Flexible (change the input composite)
- Scalable by the distance
  
- Computes rates per time, rates per recoil energy per neutrino flavor, per isotope

Fluxes can be fetched from any snwpy model

$$\frac{d^2 R}{dE_R dt_{pb}} = \sum_{\nu_\beta} N_{Xe} \int_{E_{min}^\nu} dE_\nu f_\nu(E_\nu, t, d) \frac{d\sigma}{dE_R}(E_\nu, E_R)$$

# CEvNS into

```
A = sn.Models(model_name='Forna
```

```
> Available files for this mode
```

- [0] lum\_spec\_12M\_r10000\_dat
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Fluxes can be fetched from

```
Xenon_Atom.py | Supernova_Models.py | Recoil_calculations.py | Plotter.py
65
66 Xe134 = {
67     'Type'      : 'Xe134',
68     'MassNum'   : 134,
69     'AtomicNum' : 54,
70     'Mass'      : 133.9053945,
71     'Spin'      : 0,
72     'Fraction'  : 0.104357
73 }
74
75 Xe136 = {
76     'Type'      : 'Xe136',
77     'MassNum'   : 136,
78     'AtomicNum' : 54,
79     'Mass'      : 135.907219,
80     'Spin'      : 0,
81     'Fraction'  : 0.088573
82 }
83
84 ATOM_TABLE = {
85     'Xe124' : Xe124,
86     'Xe126' : Xe126,
87     'Xe128' : Xe128,
88     'Xe129' : Xe129,
89     'Xe130' : Xe130,
90     'Xe131' : Xe131,
91     'Xe132' : Xe132,
92     'Xe134' : Xe134,
93     'Xe136' : Xe136
94 }
```

(composite)

rates per recoil energy  
pe

$$\int_{E_{min}^v} dE_v f_v(E_v, t, d) \frac{d\sigma}{dE_R}(E_v, E_r)$$

# CEvNS intro

```
A = sn.Models(model_name='Fornax_2021')
```

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Fluxes can be fetched from

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70     'Mass'      : 133.9053945,  
71     'Spin'      : 0,  
72     'Fraction'  : 0.104357  
73 }  
74  
75  
76 A = sn.Models(model_name='Fornax_2021', index=5)  
77  
78 A.compute_rates();  
79  
80 Computing for all isotopes: 100% ██████████ 9/9 [00:00<00:00, 457.87it/s]  
81  
82 100% ██████████ 9/9 [00:00<00:00, 46.67it/s]  
83  
84 100% ██████████ 9/9 [00:00<00:00, 44.24it/s]  
85  
86 Xe120 : Xe120,  
87 'Xe128' : Xe128,  
88 'Xe129' : Xe129,  
89 'Xe130' : Xe130,  
90 'Xe131' : Xe131,  
91 'Xe132' : Xe132,  
92 'Xe134' : Xe134,  
93 'Xe136' : Xe136  
94
```

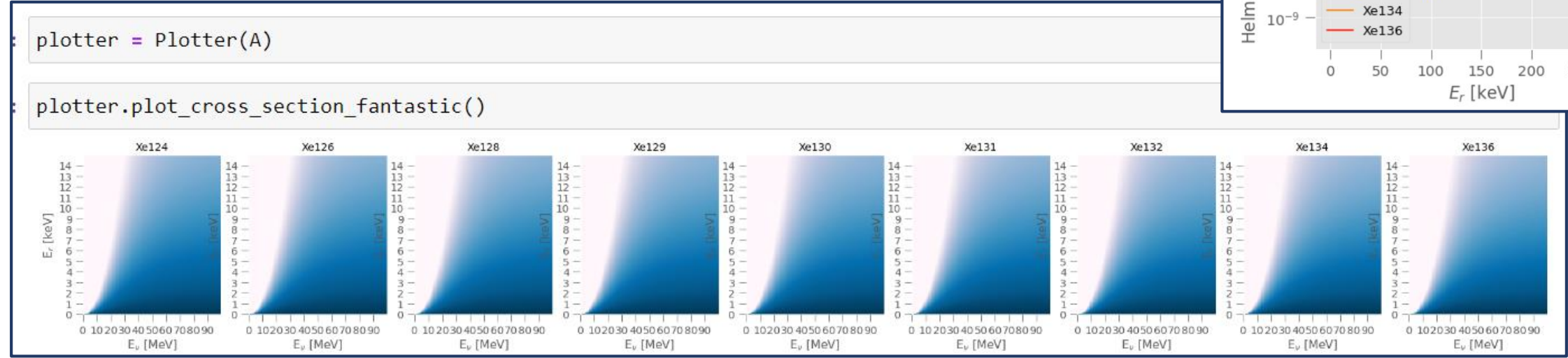
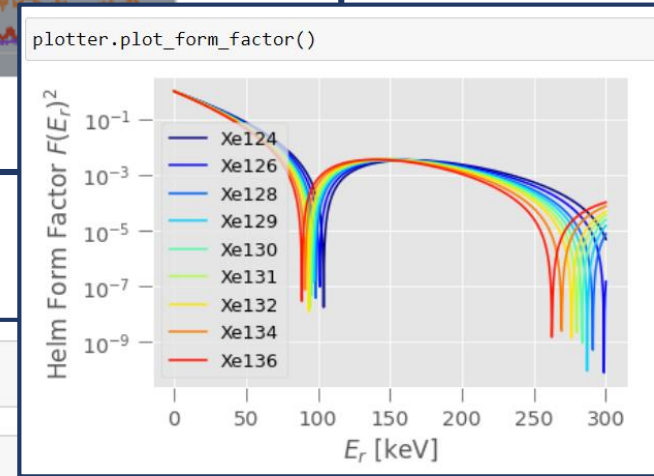
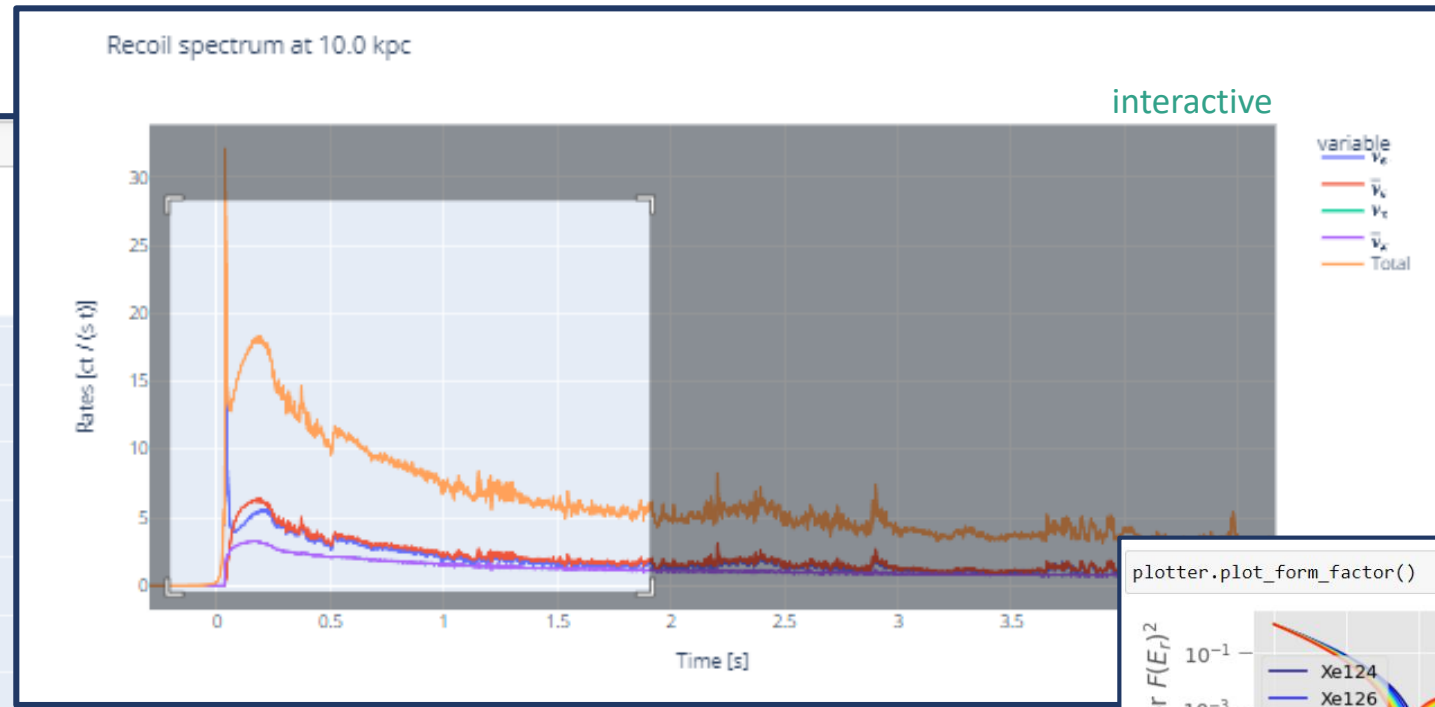
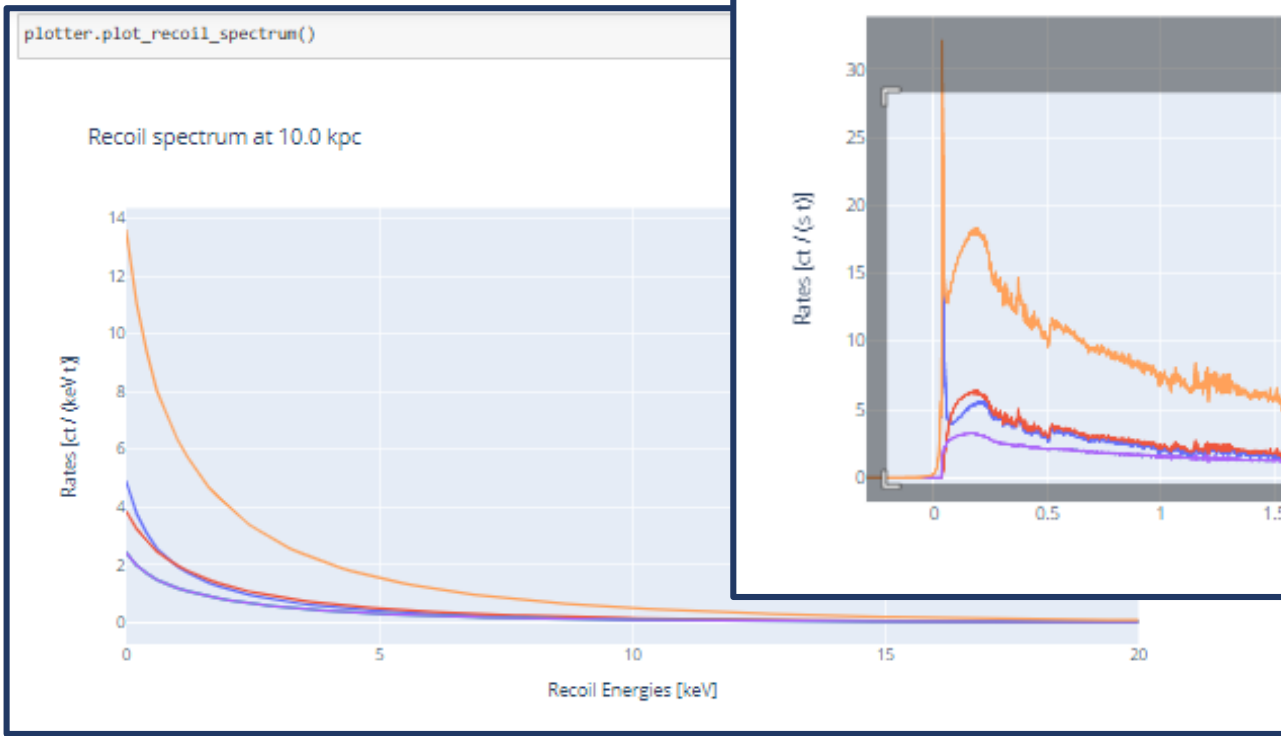
Once computed, can always be fetched in < 1s

```
A = sn.Models(model_name='Fornax_2021', index=5)  
A.compute_rates();  
Computing for all isotopes: 100% ██████████ 9/9 [00:00<00:00, 457.87it/s]  
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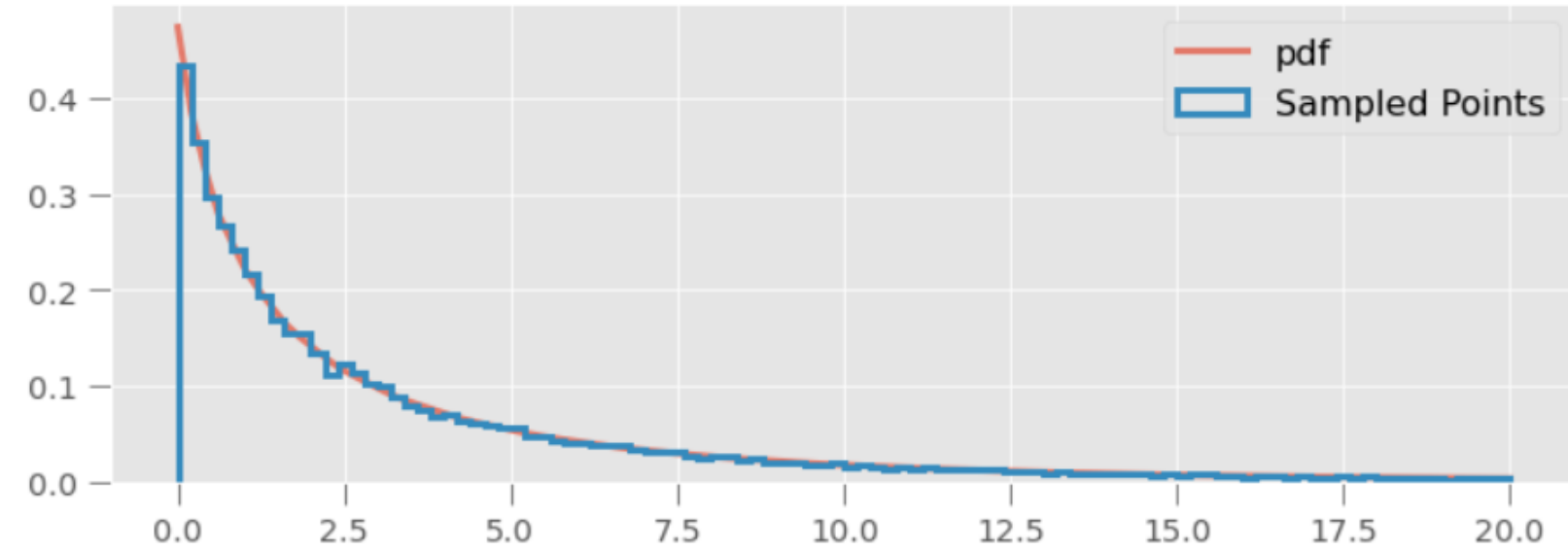
$$\int_{E_{min}^v} dE_v f_v(E_v, t, d) \frac{d\sigma}{dE_R}(E_v, E_r)$$



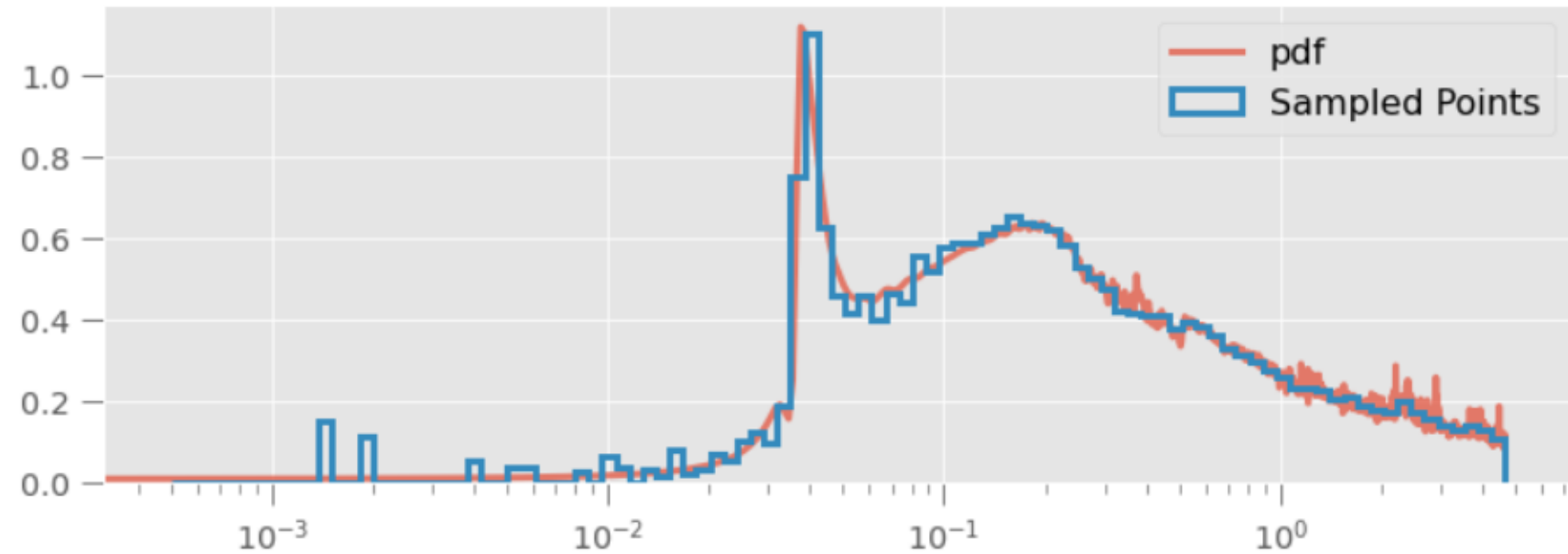
# Has a built-in plotter



## In XENONnT ...



sample recoil energies, times  
simulate and analyze signal  
using WFSim



## Summary:

- An efficient way of computing interaction rates

For any snewpy model

at any given time

for any given recoil energy

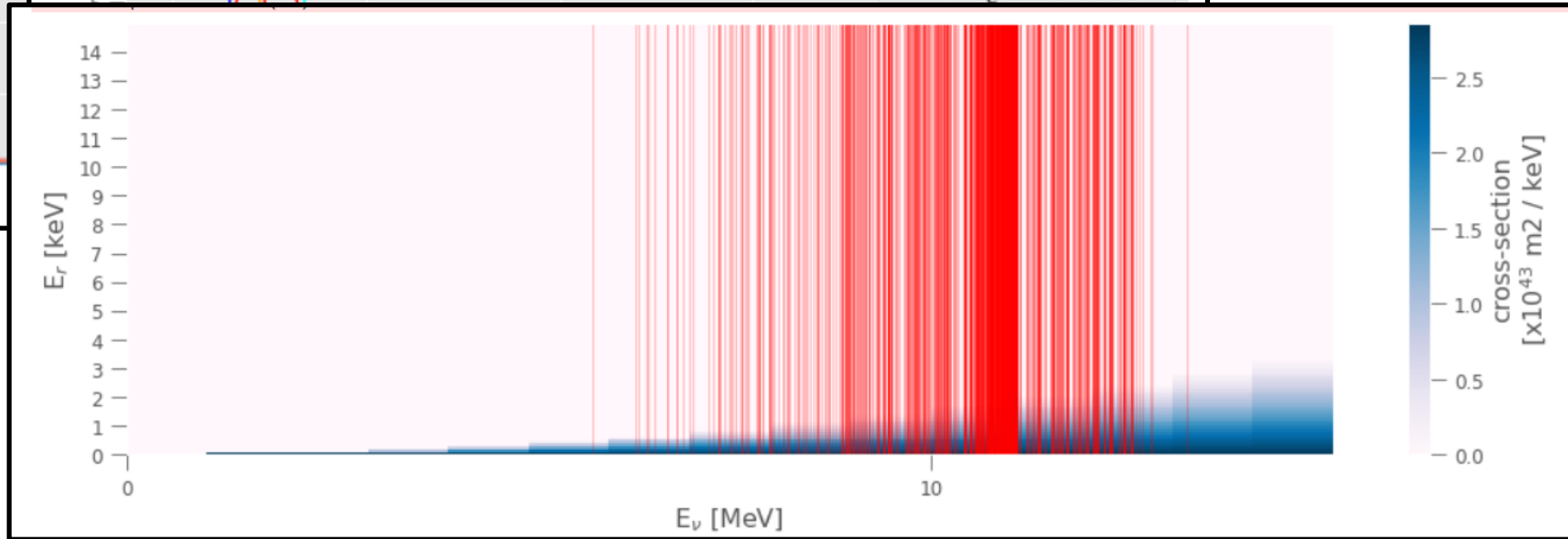
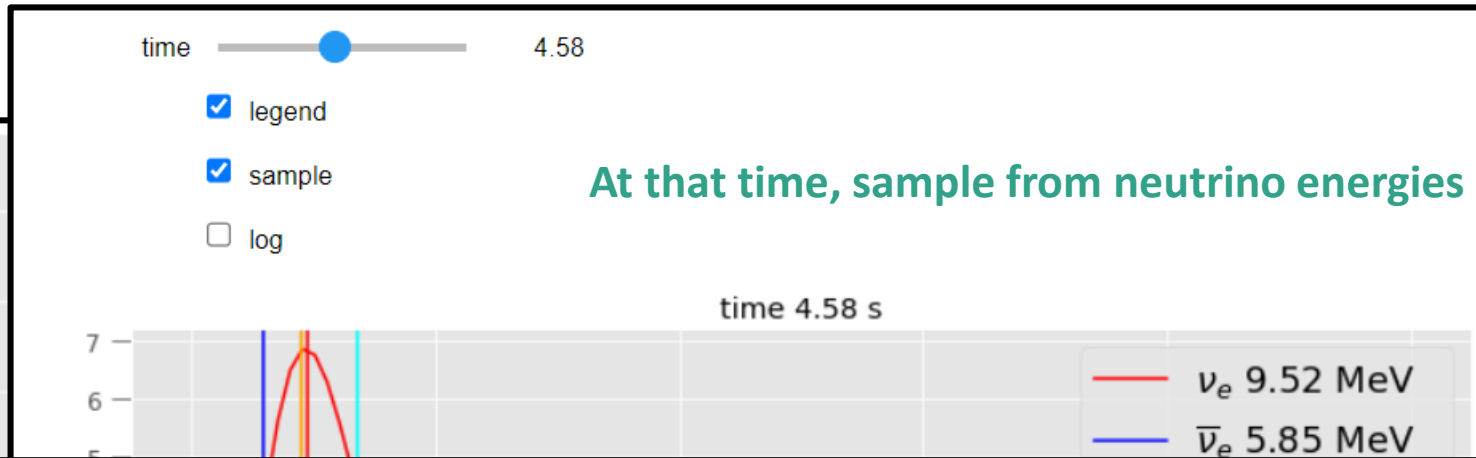
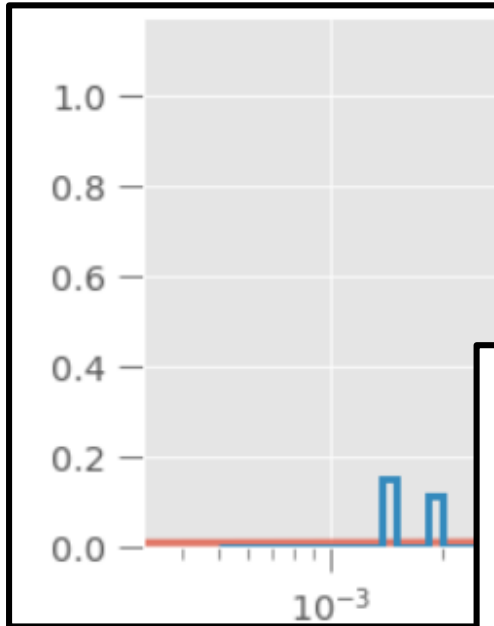
from any given neutrino energy

for any given atom/isotope

for all neutrino flavors

**Backup**

# In XENONnT ...



At that neutrino energy sample a recoil energy based on cross-section probability

This method actually gives you a terrible spectra  
Regardless if you sample from time first (almost only returns first second)  
Or sample time uniformly and look at the energy distributions.

Although you get those neutrinos more, they don't recoil energetic enough.  
So you only see the high energy ones (and there are still a lot  $10^{12}$ )

It makes more sense to study what we can see, and energy neutrino can detect those.  
Then looking into flux distribution to see what models can generate enough of those events, and at what times

Sampling from time integrated recoil energy distribution is probably the best approximation to this.