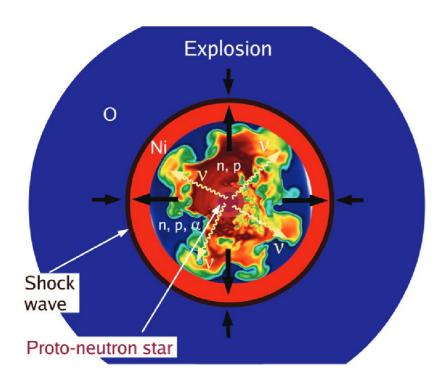
Unfolding of the supernova burst neutrino spectrum

Thilo Birkenfeld Physics Institute III B, RWTH Aachen University Europe Meeting, 24.10.2022





Core-Collapse Supernova



- Core collapse supernovae produce a short (~10s), strong (L~10⁵³ erg/s) neutrino burst of all flavor
- Neutronization

$$e^- + p \to n + \nu_e$$

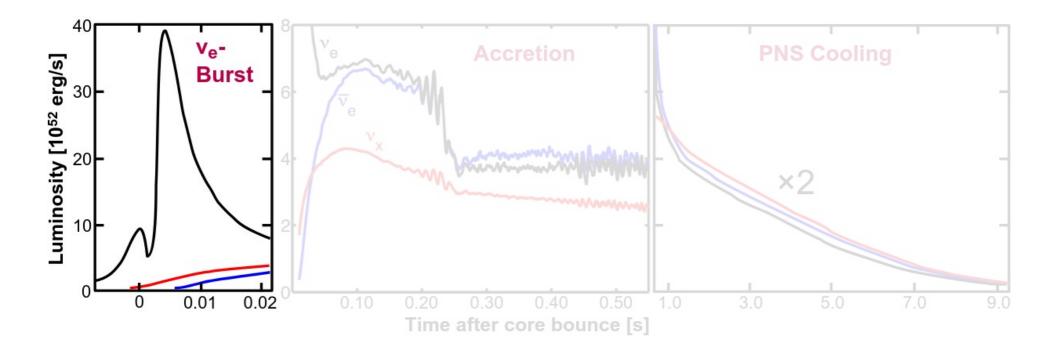
- Pair-production $e^- + e^+ \rightarrow \nu_f + \bar{\nu}_f$
- And more ...
- $\langle E_{\nu} \rangle \sim 10 15 \,\mathrm{MeV}$





Neutronization burst



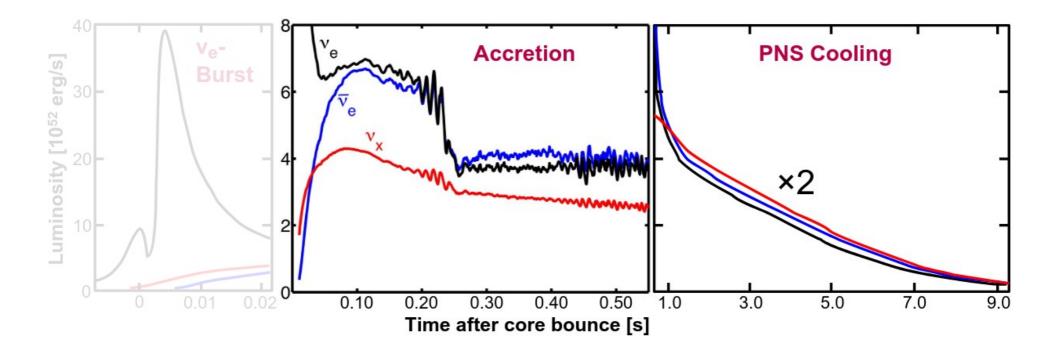


- Within ~ ms a neutronization burst emits a large number of electron neutrinos
 - Caused by shock wave becoming transparent for neutrinos
 - Highly non-thermal spectrum



Accretion and cooling





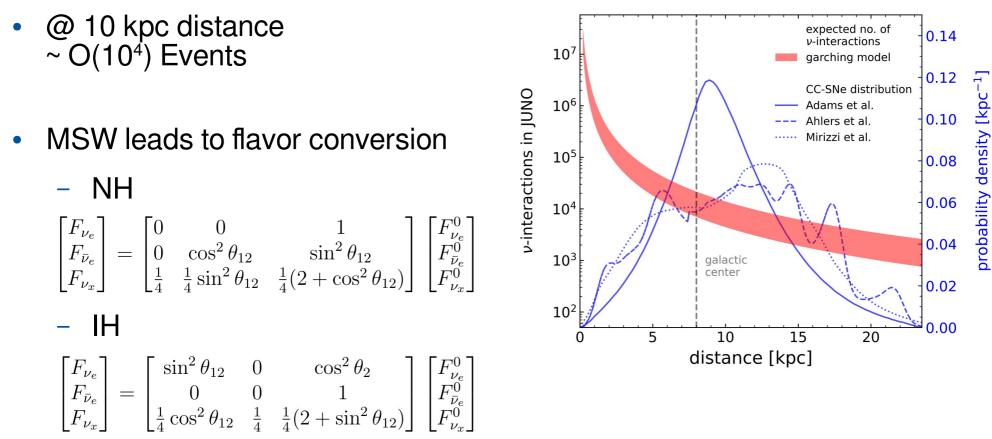
- Accretion: Luminosity and $\langle E_{\nu} \rangle$ rapidly changing
- Cooling phase: thermal spectra with decreasing temperature







Cumulative Spectra

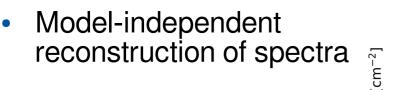


- Time integrated v_f-spectra are mixture of different compositions (thermal and non-thermal)
 - Dependence on details of the supernova

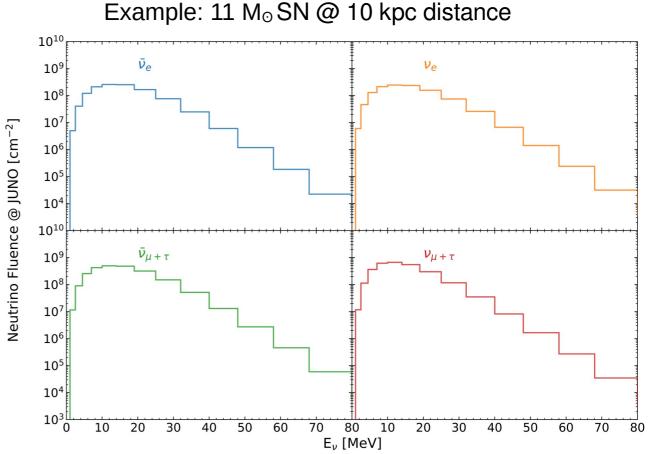


Motivation





 \rightarrow Unfolding







Event Typology

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Supernova Neutrinos in JUNO

- Multiple channels
 - Inverse Beta Decay (IBD)
 - e⁺ + n, Δt ~ 220 μs
 - Proton elastic scattering (PES)
 - Electron elastic scattering (EES)
 - ¹²C interactions
 - ${}^{12}C^*$ excited state $\rightarrow \gamma$ (15.1 MeV)
 - ${}^{12}N/{}^{12}B$ production - $e^{-(+)} + e^{+(-)}$, $\Delta t \sim 10 \, \text{ms}$

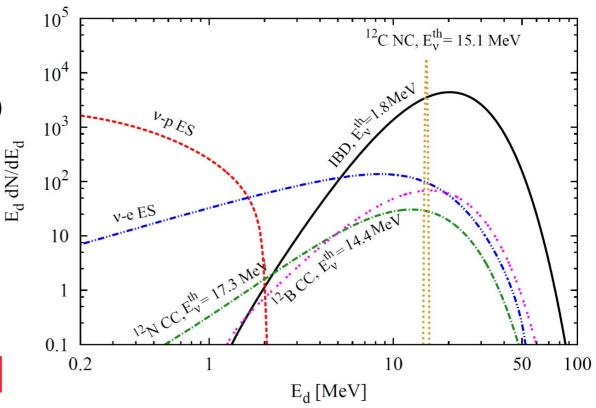


Fig 1. Expected energy spectrum for 10 kpc corecollapse Supernova

$$\sigma_f^{channel} \neq \sigma_{f'}^{channel}$$
 in general

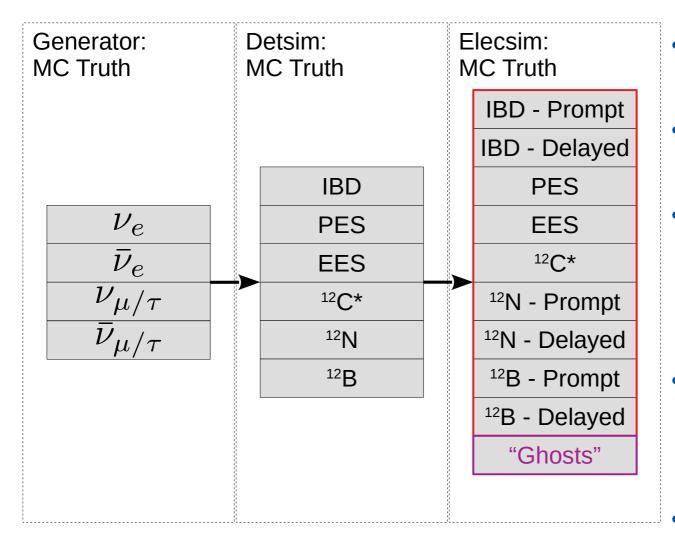
Channel identification required for multi flavor analysis



MC Truth



• What is MC truth type information? (Without cluster algorithm)



- Timing
- Secondary particle type
- Each readout window will be assigned to biggest contributor (1 MC Truth)
- "Ghosts" from afterpulses or splitted events
 - For event selection





Event Selection

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• 48 CC-SN @ 10 kpc

dags	gar81121-97	gar81123-102	gar82500-107	gar82502-112	gar82700-117	gar82701-122	gar82703-127	gar84001-132	gar84002-137
gar81120-93	gar81121-98	gar81123-103	gar82501-108	gar82502-113	gar82700-118	gar82702-123	gar82703-128	gar84001-133	gar84003-138
gar81120-94	gar81122-100	gar81123-104	gar82501-109	gar82503-114	gar82700-119	gar82702-124	gar84000-129	gar84001-134	gar84003-139
gar81120-95	gar81122-101	gar82500-105	gar82501-110	gar82503-115	gar82701-120	gar82702-125	gar84000-130	gar84002-135	gar84003-140
gar81121-96	gar81122-99	gar82500-106	gar82502-111	gar82503-116	gar82701-121	gar82703-126	gar84000-131	gar84002-136	gdml.root

- 3 different seed for each model
- ~ 500'000 events
- Full Chain (including elecsim)





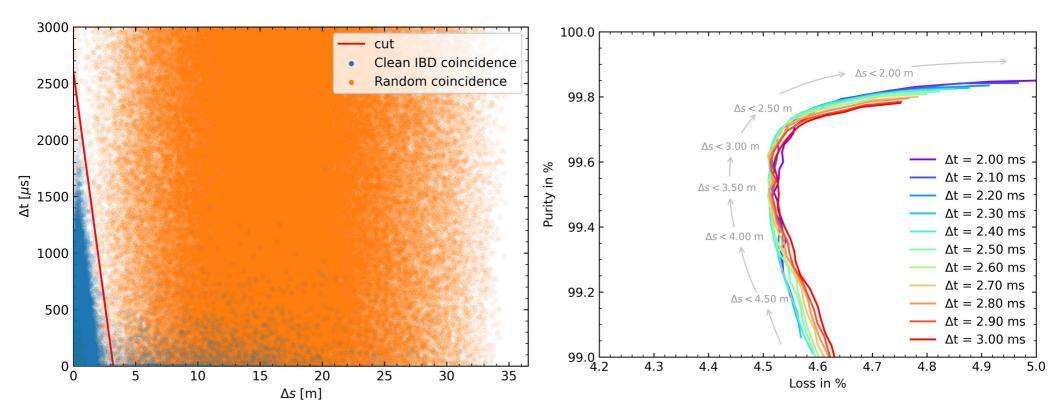
IBD selection



- Delayed energy: E_{max} < 5.25 MeV
- Prompt energy: E_{min} > 1.0 MeV

Purity = # correct selection / # total selection Loss = # missed events / # all events

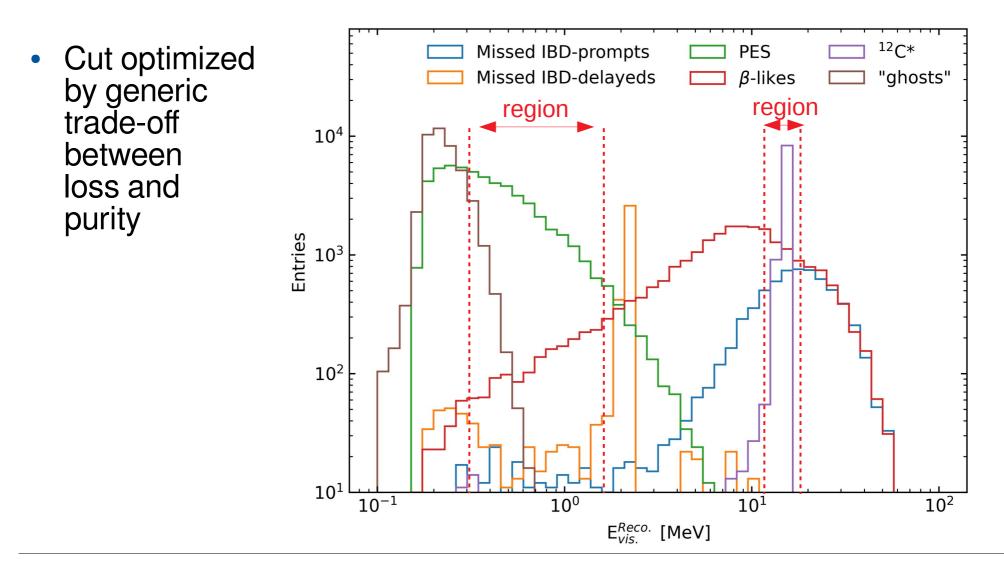
• Optimized by minimal loss \rightarrow (loss will contaminate other channels)





Distribution after IBD Selection

Select PES and ¹²C* events from energy region



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- "ghosts" are in lowest energy region
 - Everything below min. proton energy regarded as "ghosts"

- Everything is selected except β-likes
 - Left over is β-like channel





Spectra Unfolding

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Response Matrix Construction



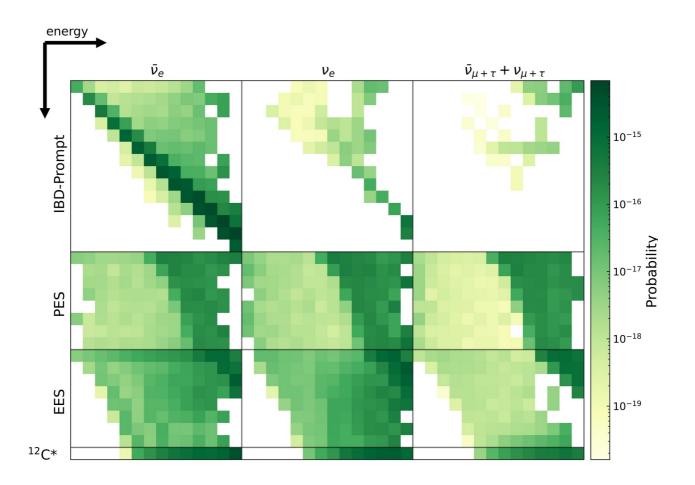
- With MC Truth definition Neutrino Event ν_f, E_{ν} ch, E_{vis} . 1. With neutrino flavor simulation ν_f, E_{ν} ch, E_{vis} . 2. ν_f, E_{ν} ch, E_{vis} . 3. Get bin IDs: E_{ν} $u_f,$ 3 i =ch, $E_{vis.}$ 3 3 24 $\mathbf{0}$ ()()6 450 $\left(\right)$ \mathbf{O} j = 33 21 + 11 0 0 0 Add each event to transition matrix: 0 126 $\mathbf{2}$ 0 () 3 0 5 4 0 0
- Normalize each column by no. v's that pass through JUNO
 probabilities (M)

Direct Unfolding



Response Matrix





• Important:

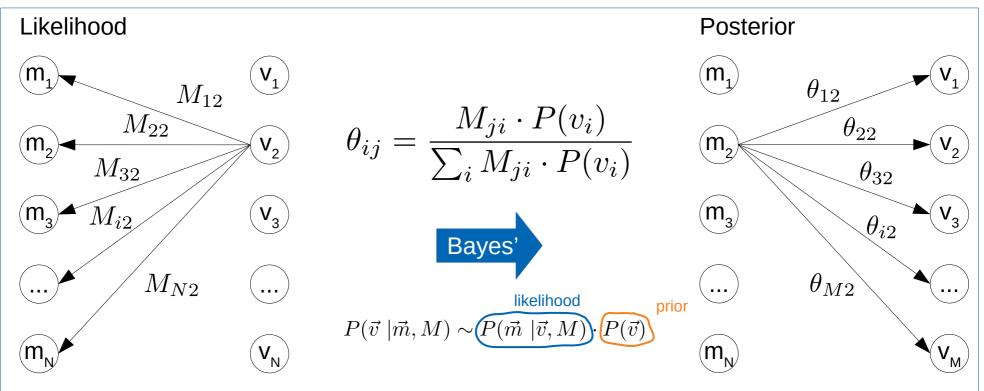
Physical spectrum is original neutrino energy distribution

• Includes cross-section, trigger and selection efficiencies and energy smearing



Inference Step





Full physical spectrum:

$$\tilde{v}_i \approx \frac{1}{\epsilon_i} \sum_{j=1}^{n_{\vec{m}}} \theta_{ij} \cdot m_j$$

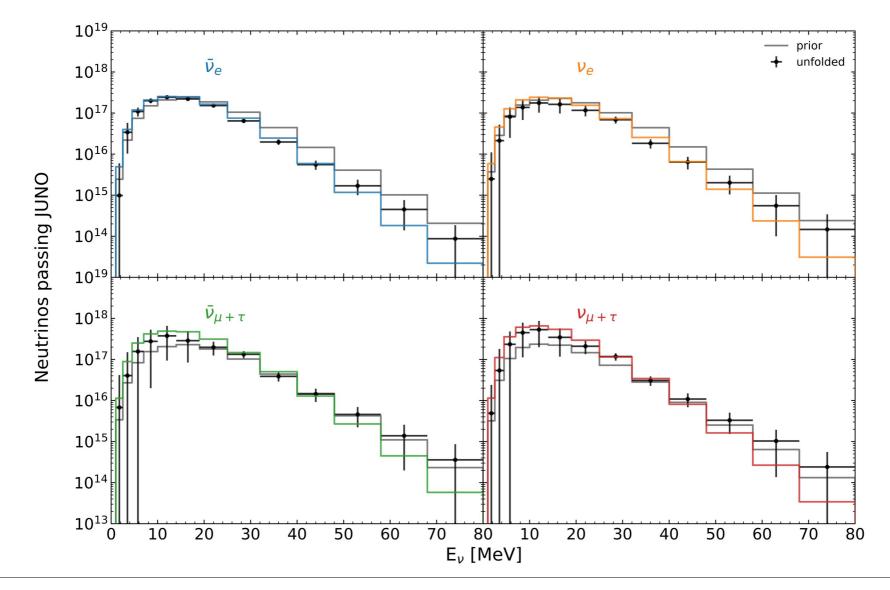
$$\epsilon_i = \text{efficiency}$$

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Example Unfolding





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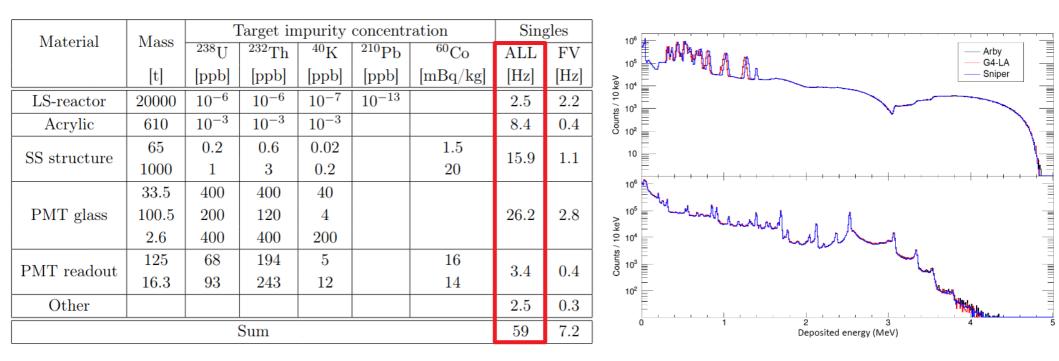
Backgrounds

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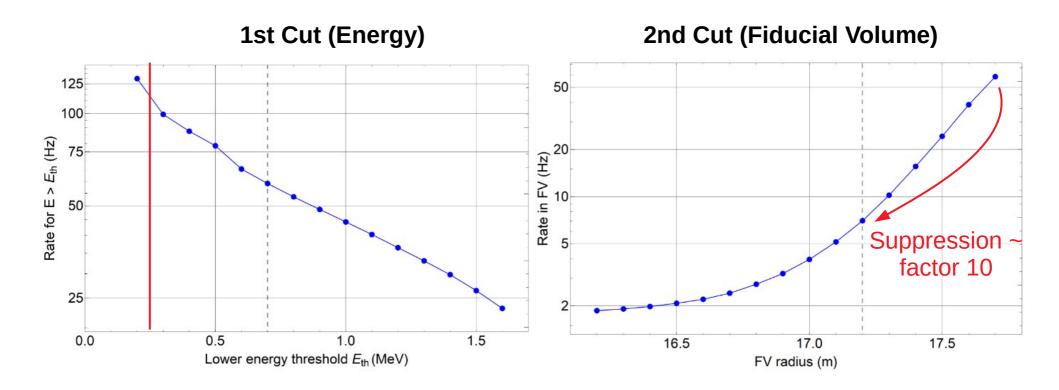
https://juno.ihep.ac.cn/cgi-bin/Dev_DocDB/ShowDocument?docid=7057



- Without any cuts ~ 590 singles in triggered (~ 10 second SN signal)
- BUT: This is with a Energy threshold of 700 keV







- With my energy cut and FV cut \rightarrow ~110 events in 10 secs.
- But loose ~ 4% data (with vertex reco.)

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Background selection



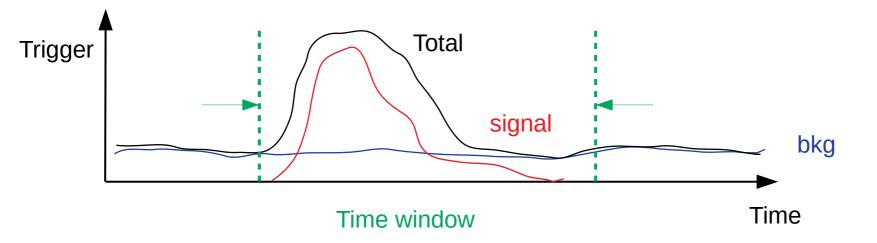
- ~ 110 with FV cut (10 second)
- In low energy region:
 - Contaminates PES
 - Contaminates EES
- More sophisticated selection should improve selection
 - Pulse Shape discrimination for PES
 - Characteristic energies for certain decays



Include in Simulation



Again timing is crucial for SN



- Time window (start, end) should be optimized for each SN (or any bursts)
 - Different duration + different distance gives different signal to bkg ratio
- There will be one or two online burst alerts from global trigger and MM trigger
 - Fixed length (60 secs)
 - Offline analysis might yield better results
- This effect needs to be simulated properly \rightarrow to be implemented

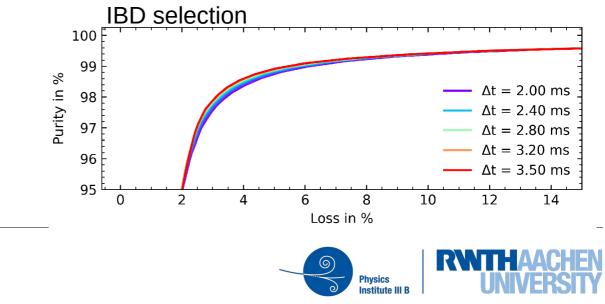




Summary / Outlook



- Established working unfolding scheme
 - Direct unfolding (regularization on neutrino spectra)
- Primary results very promising
- Estimated small contribution from radioactive backgrounds
- Produced larger data set ~ 55 models with more variety & updated JUNO simulation



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Backups

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Reformulated inference



Define: $\theta_{ij} = \frac{M_{ji} \cdot P(v_i)}{\sum_i M_{ji} \cdot P(v_i)}$

$$\tilde{v}_i \approx \frac{1}{\epsilon_i} \sum_{j=1}^{n_{\vec{m}}} \theta_{ij} \cdot m_j$$

- \tilde{v}_i reconstructed bin content in physical distribution (collection of \tilde{v}_i gives reconstructed distribution)
- ϵ_i in inefficiency of physical bin i
- $n_{ec{m}}$ number of bins in measured distribution





Response construction



- With MC simulations we get the physical spectrum \vec{v} and the observed spectrum \vec{m}
 - In our case we actually have a collection of simulations and spectra
- Using the MC truth to measured relations on event level we can construct our response matrix
 - Which is an **estimator** for the true response matrix
 - Statistical fluctuations from limited amount of MC
 - Systematics coming from the model choices

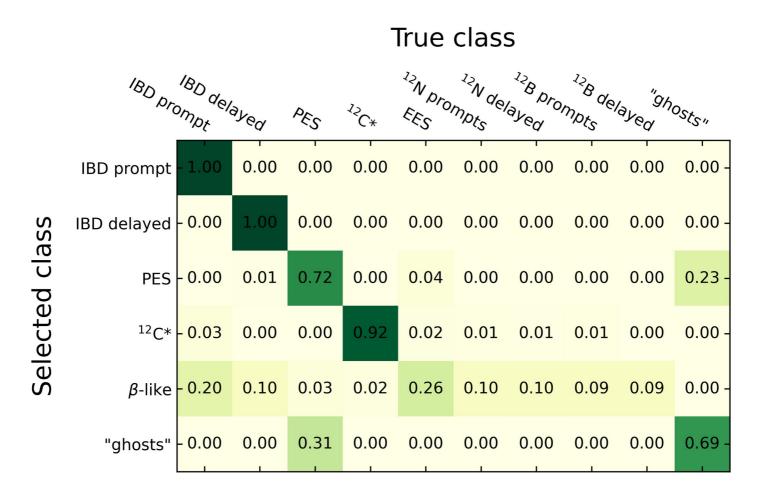
 $\overline{M}_{MC}\approx M$





Performance – Contamination Matrix





 Entries = # events selected / # events in selected class (fraction of selected events from true class...)

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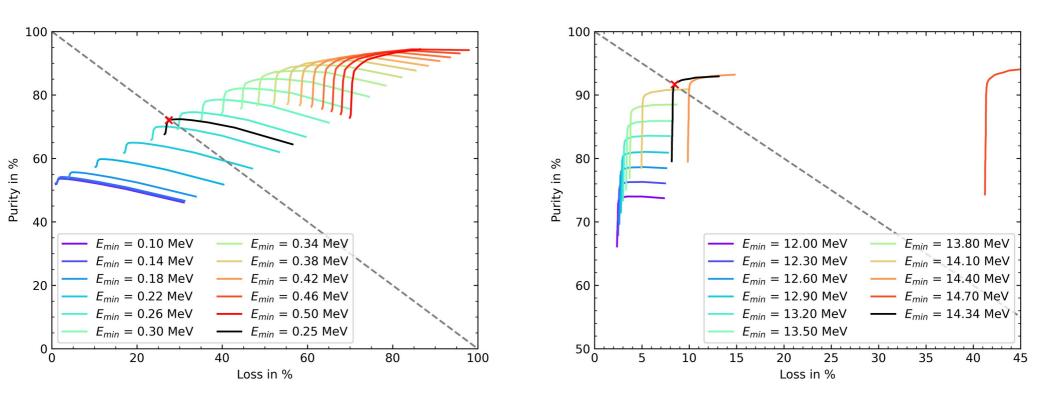


PES & ¹²C* - Optimization



PES

¹²C*



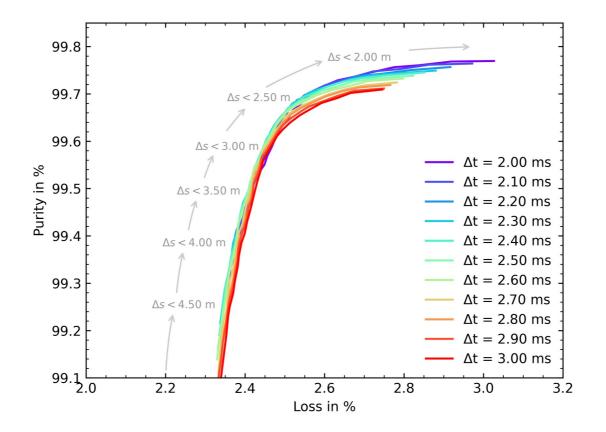
- Gray line is diagonal between max purity and max loss
- Define optimum as interception of diagonal with highest purity

Generic trade-off between purity and efficiency



IBD delayed



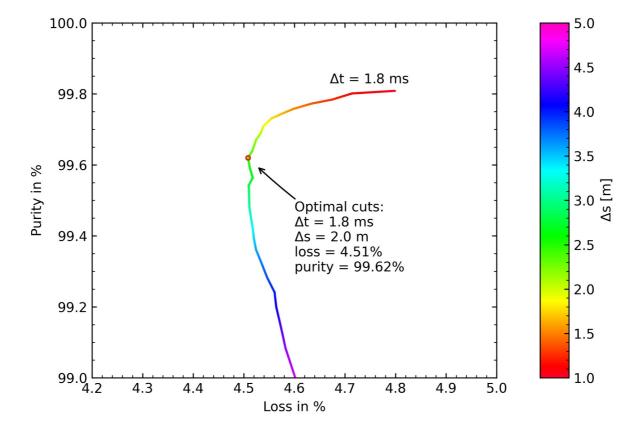




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IBD prompt optimal curve



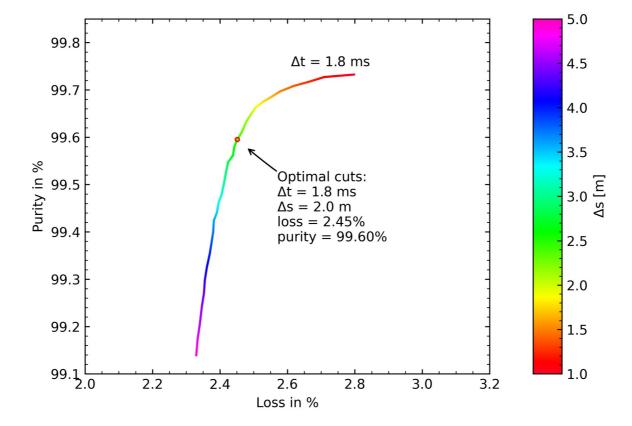




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IBD delayed optimal curve







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Uncertainty handling – statistical



- Original d'Agostini method used gaussian error propagation
 - Did not take poisson distribution into account
 - Did not take error propagation from updated priors into account

Does **not** describe cases with **small numbers** well!

- Improved method (Improved iterative Bayesian unfolding)
 - Added poisson fluctuations
 - Estimates uncertainties from sampling underlying distributions

$$M_{sample}, m_{sample}, (\vec{v}'_{sample}) \rightarrow \vec{v}'_{\vec{v}'_{sample}} \rightarrow \overline{\vec{v}'}, Cov_{ij}$$

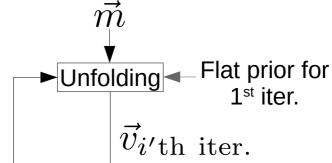


Bayesian inference and prior update

- Starting from a flat prior, we basically updated our knowledge (in this case no knowledge) of the physical spectrum with our unfolding
- Now that we have an "updated" version of the physical distribution \vec{v}' we can rerun our unfolding by assuming the shape of \vec{v}' as prior:

 $\rightarrow \vec{v}''$ 2nd iteration

- This procedure can be redone iteratively until the updated distribution converges (*Iterative bayesian prior update*)
 - To much iterations blow up statistical uncertainties
 - In some cases the unfolded distribution does not converge



- Starting from a **non-flat prior** (prior knowledge), usually the physical distribution \vec{v}' is only updated once
 - Multiple iterations can lose initial prior knowledge





Bayesian inference and prior update



- Starting from a **flat prior**, we basically updated our knowledge (in this case no knowledge) of the physical spectrum with our unfolding
- Now that we have an "updated" version of the physical distribution $ec{v}'$ we can rerup our unfolding by assuming the shape of \vec{v}' as prior The use of priors acts as **regularization** for this method. This pro converges (as smooth priors are usually assumed) (Iterative Tom Other unfolding methods (e.g. SVD unfolding) stati use manual regularizations or for to constrain the blow up of statistical fluctuations. er. In sc distr
- Starting from a **non-flat prior** (prior knowledge), usually the physical distribution \vec{v}' is only updated once
 - Multiple iterations can lose initial prior knowledge



Pile-Up



Readout event:

detector

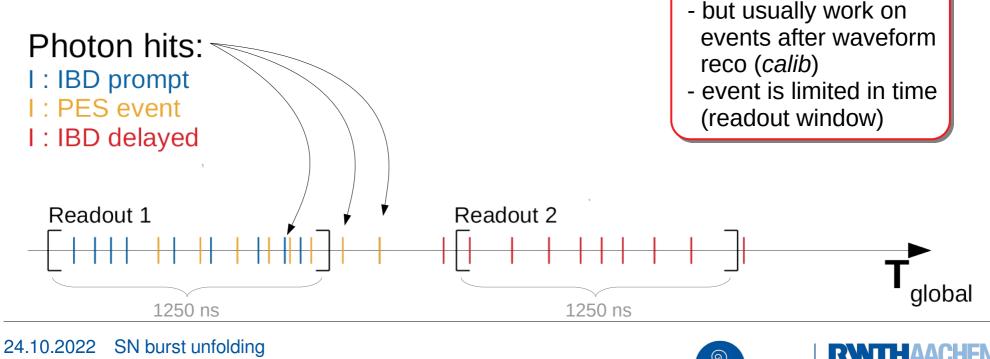
elecsim

Physics Institute III B

- "raw" data output of

- in software given by

- Photon hits from N neutrino events (usual just two) in one readout window (after *elecsim*)
- Low signal rate:
 - Pile-up of signal + background
- High signal rate:
 - Pile-up of signal + signal

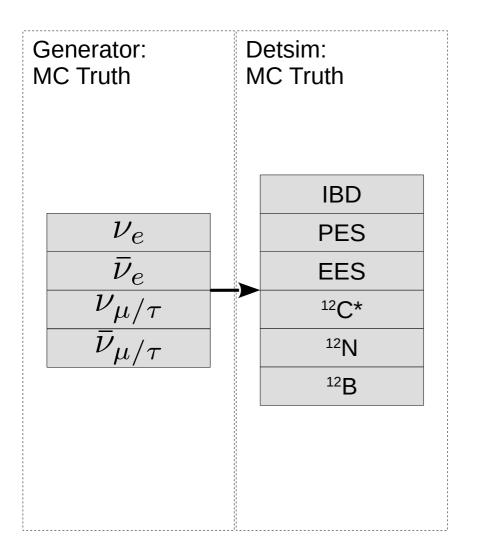


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MC Truth



• What is MC truth type information?



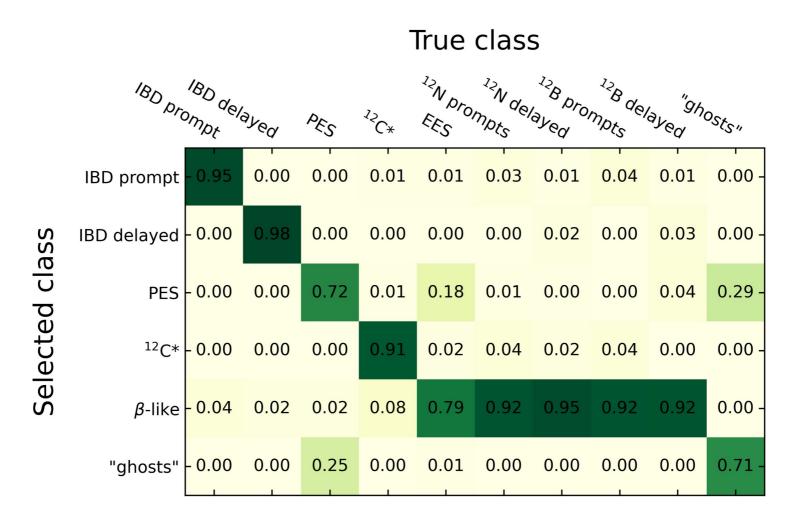
- Interaction channel
- Secondary particle type
- Secondary energy
- Secondary position
- For event reconstruction (energy + vertex)





Total Confusion Matrix





 Entries = # events selected / # events in true class (fraction of true events selected into class...)

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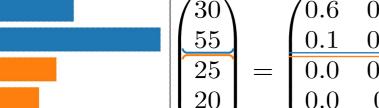


For binned distributions problem can be described by matrix multiplication

Unfolding Multiple Spectra

$$\vec{m} = M \cdot \vec{v}$$

Physical distribution



12

Measured distribution

	(0.6)	0.05	0.1	0.0	0.0	0.0
	0.1	0.75	$\begin{array}{c c} 0.1\\ 0.2 \end{array}$	0.0	0.0	0.0
=	0.0	0.05	0.7	0.05	0.0	0.0
	0.0	0.0	0.00	0.6	0.6	0.4
	$\left(0.0 \right)$	0.0	$0.7 \\ 0.00 \\ 0.00$	0.25	0.4	0.6/
	`					,

- Can concatenate multiple physical distributions (e.g. neutrino flavor)
- Can concatenate multiple measured distributions (e.g. interaction channels)
- Response consist of different blocks
 - Block describes bin migration from one physical class to one measured class



40

60

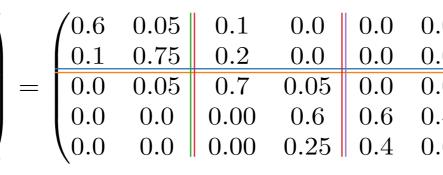
30

20

10

5





Inference problem



- Now consider a true measurement (or a different MC sample)
- Simple matrix inversion (M^{-1}) not possible (response might not be invertible, but even if, the inversion only holds for expectation values **not for sample measurement**)

$\begin{pmatrix} 25\\58\\21\\25\\7 \end{pmatrix} =$	$\begin{pmatrix} 0.6 \\ 0.1 \\ 0.0 \\ 0.0 \\ 0.0 \\ 0.3 \end{pmatrix}$	$\begin{array}{c} 0.05 \\ 0.75 \\ 0.05 \\ 0.0 \\ 0.0 \\ 0.0 \\ 0.15 \end{array}$	$\begin{array}{c} 0.1 \\ 0.2 \\ 0.7 \\ 0.00 \\ 0.00 \\ 0.00 \end{array}$	$\begin{array}{c} 0.0 \\ 0.0 \\ 0.05 \\ 0.6 \\ 0.25 \\ 0.1 \end{array}$	$0.0 \\ 0.0 \\ 0.0 \\ 0.6 \\ 0.4 \\ 0.0$	$\begin{array}{c} 0.0 \\ 0.0 \\ 0.0 \\ 0.4 \\ 0.6 \\ 0.0 \end{array}$	(? ? ? ? ? ?
	0.3	0.10	0.00	U.L	0.0	0.0 🔸	

- How can we infer the physical spectrum using our knowledge on the response?
 - Keep in mind we don't know the true number of lost events (inefficiency bin), but we still know the mean inefficiency
- Bayes' theorem can help: $P(\vec{v} \mid \vec{m}, M) \sim P(\vec{m} \mid \vec{v}, M) \cdot P(\vec{v})$ likelihood
 prior
- Method of d'Agostini Improved iterative Bayesian unfolding

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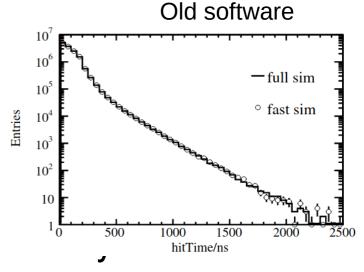


How about muons?



overburden	Muon rate	Average energy	Yellow Book veto live time	New veto(doc 5837)
~748 m (w/o hall shape, CD center)	0.00303 Hz/m ²	215 GeV	83%	94.7% (estimated)
~700 m (w/o hall shape, 50 m up, doc-2512)	0.00367 Hz/m ²	209 GeV	79% (estimated)	93.6%
New ~644.1 m (2020.11, Jilei Xu, w/ hall shape doc-6399)	0.0041 Hz/m ²	207 GeV	77% (estimated)	91.6% (JUNO-doc- 5837)

- With detector cross-section $\rightarrow \sim 4$ Hz muon rate
- Muon pulse ~ 1.5 to 2 μs
- Small fraction of downtime
 - 10 kpc SN: pile up with signal very





Backgrounds considerations



- How big is actual bkg contribution (with global trigger >200 MeV)?
 - New paper about to be published

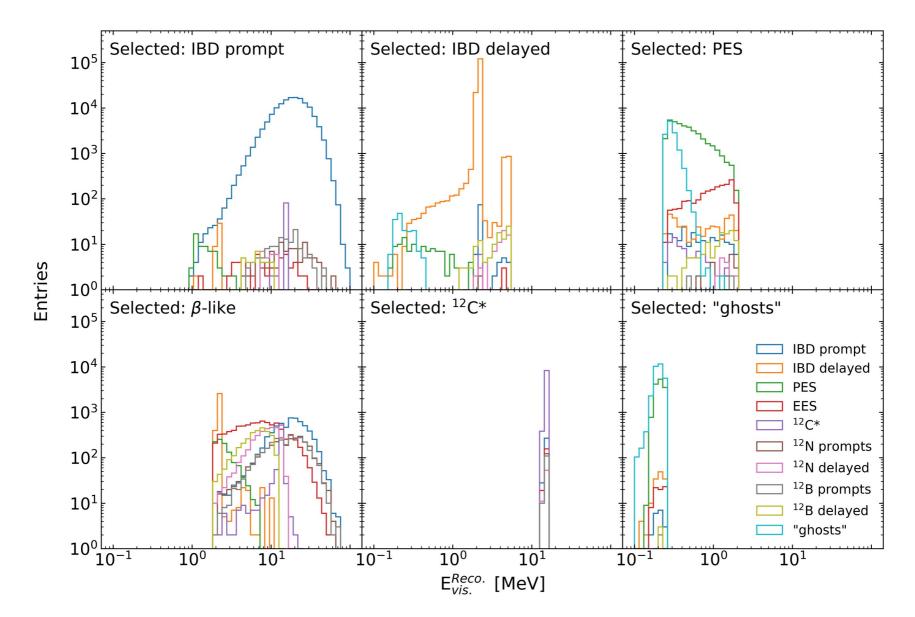
Selection	IBD efficiency (%)	IBD	Geo-vs	Accidental	⁹ Li ⁸ He	Fast n	(α,n)
-	-	59	1.5	$\sim 2.2 \times 10^4$	71	-	-
Fiducia volumel	91.5						
Energy cut	99.83	52	1.3	0.66		0.1	0.05
Time cut	99.02	53					
Vertex cut	99.23						
Muon veto	91.6	48	1.2	0.8	1.4		
Combined	82	⁸² 48 3.55					
Updated IBD efficiency and background rate.							

- In readout of ~ 10 sec. reactor, Geo-vs & ⁹Li. ⁸He neglectable
- How about accidental? (here accidental coincidences)



Selected Energy Distributions





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