

Core-collapse Supernova Model Discrimination

Jost Migenda
they/them

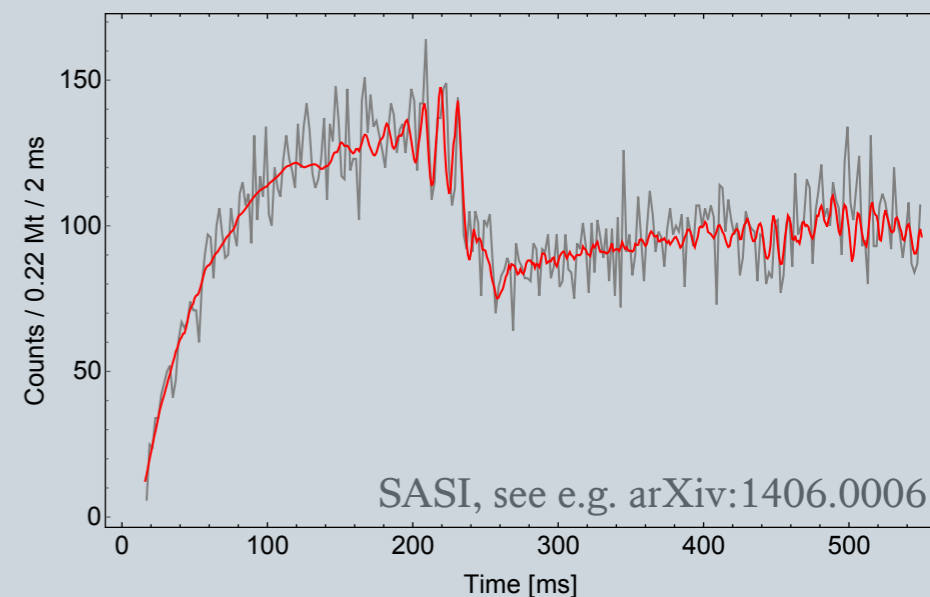
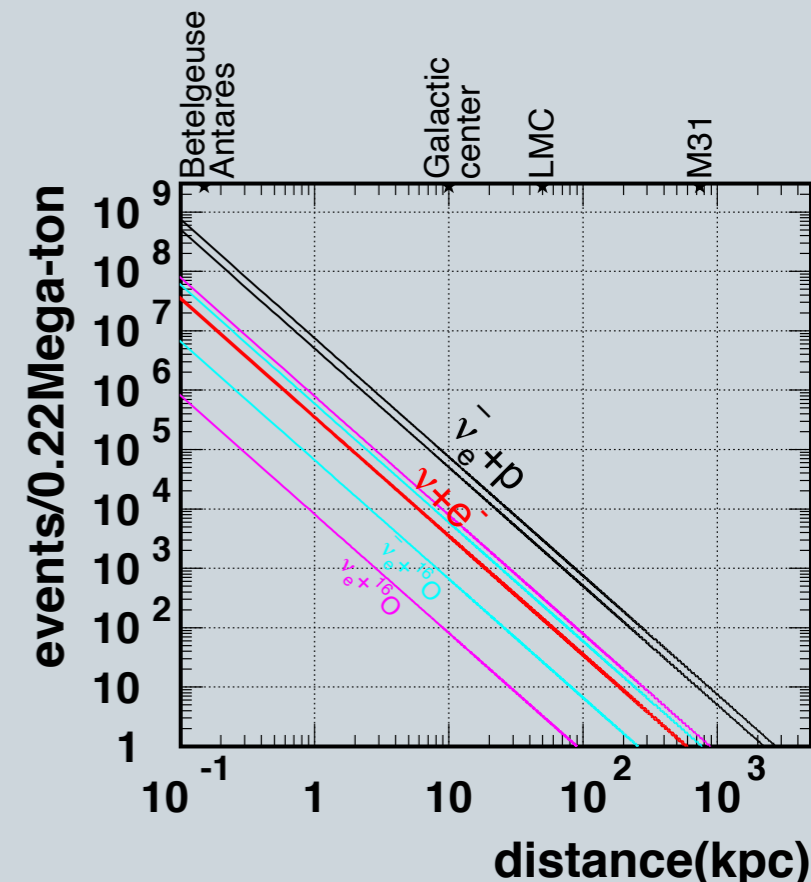
based on **arXiv:2002.01649**
and **arXiv:2101.05269**

... in Hyper-Kamiokande

based on **arXiv:2002.01649**
and **arXiv:2101.05269**

Core-collapse Supernova Observations with Hyper-Kamiokande

- ♦ HK offers both large statistics and event-by-event energy information
 - ♦ Order of magnitude larger than Super-K, DUNE, JUNO
 - ♦ IceCube: more events, but no energy information for individual events
- ♦ 54k–90k events for SN at 10 kpc
 - ♦ ~3k events for SN in LMC
- ♦ Directionality: $\sim 1^\circ$ (via ν_e -scattering)
- ♦ Most sensitive to $\bar{\nu}_e$

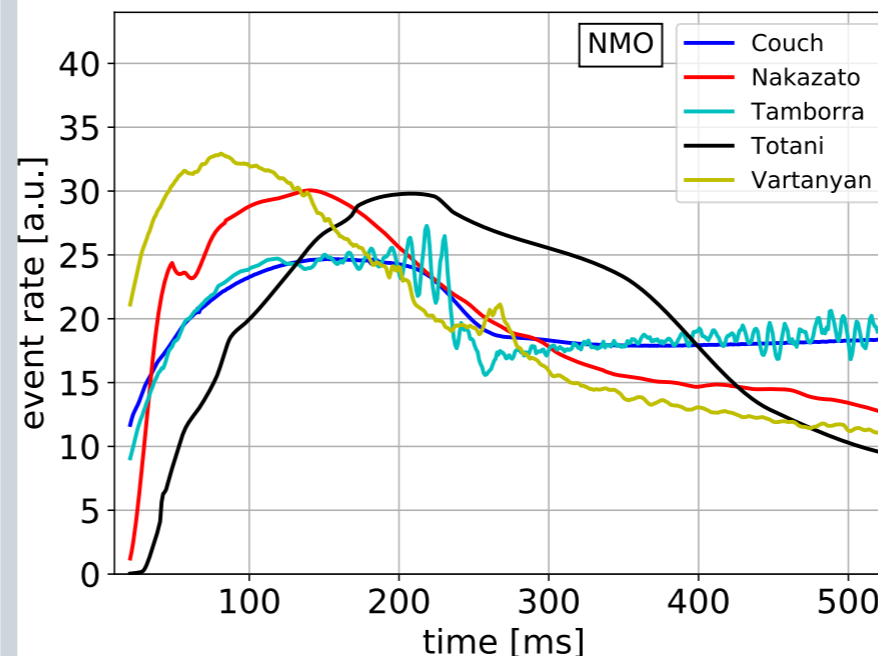
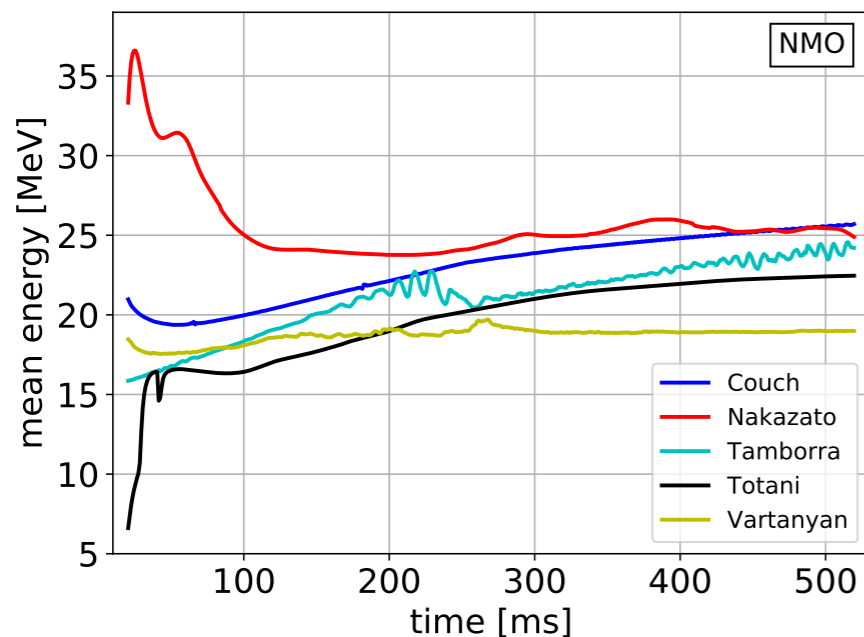


Supernova Models

Models by different groups, using various approximations

→ telling models apart can help understand the explosion mechanism

Model	Mass		events at 10 kpc*	N=100	N=300
Totani arXiv:astro-ph/9710203	20 M _⊙	1D	19716	140 kpc	81 kpc
Nakazato arXiv:1210.6841	20 M _⊙	1D	17978	134 kpc	77 kpc
Couch arXiv:1902.01340	20 M _⊙	1D	27539	166 kpc	96 kpc
Vartanyan similar to arXiv:1804.00689	9 M _⊙	2D	10372	102 kpc	59 kpc
Tamborra arXiv:1406.0006	27 M _⊙	3D	25021	158 kpc	91 kpc



* during 20–520ms after core bounce, assuming Normal Ordering

sntools: A Supernova Event Generator

- ♦ Modern & most precise cross-sections for main interaction channels
- ♦ Modular & easily extensible
 - ♦ Hundreds of SN simulations by different groups (data publicly available via SNEWPY)
 - ♦ 8 flavour transforms between SN and detector
 - ♦ Water, liquid scintillator & water-based LS
 - ♦ Adopted by HK, WATCHMAN, THEIA, SNO+
- ♦ Open source: <https://github.com/JostMigenda/sntools> & published in JOSS ([DOI:10.21105/joss.02877](https://doi.org/10.21105/joss.02877))

Data Sets

- ♦ Used sntools to generate 1000 data sets each for
 - ♦ 5 different SN models
 - ♦ Normal & inverted mass ordering
 - ♦ $N=100, 300$ events per data set
- ♦ Consider 20–520 ms post bounce only
 - ♦ Accretion phase is most interesting physically (late times: PNS cooling, similar across models)
 - ♦ Can include advanced 3D models, where computing time limitations only allow simulating <1 s

Analysis

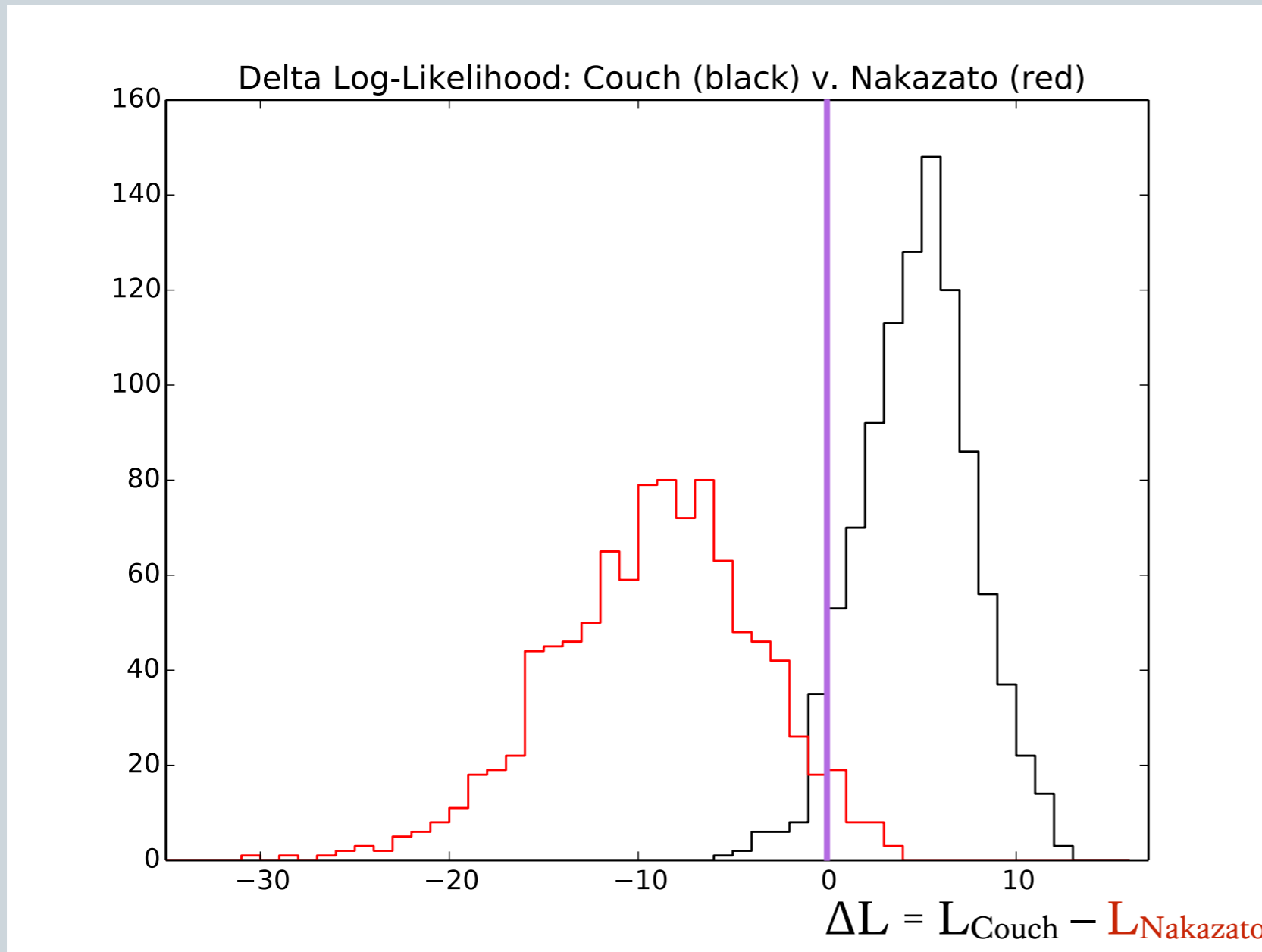
- ♦ Ran full detector simulation & reconstruction for all data sets
- ♦ Applied cuts → effectively background-free in 500 ms interval
 - ♦ $E_{\text{reco}} > 5 \text{ MeV}$ (eliminate low-E backgrounds)
 - ♦ Vertex inside fiducial volume (avoid higher backgrounds & worse reconstruction near walls)
- Of 100 (300) events, 80–85% remain after trigger and cuts
- ♦ Per data set: calculate unbinned log-likelihood L for each SN model
 - ♦ Based on Loredo & Lamb, *Annals N. Y. Acad. Sci.* 571 (1989) p. 601–630
 - ♦ Extended to include multiple interaction channels

$$L = \ln \mathcal{L} = \sum_{\text{evt } i} \ln \left(\frac{d^2 N(t_i, E_i)}{dt dE} \right)$$

Model Comparison

Normal ordering
N=100 evt/dataset

→ Use $\Delta L = L_A - L_B$ to determine whether model A or B better describes any given data set



Data sets generated
from Nakazato model

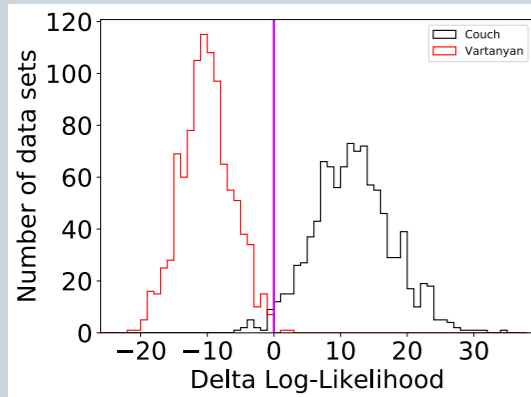
Data sets generated
from Couch model

→ Good model separation with just N=100 events

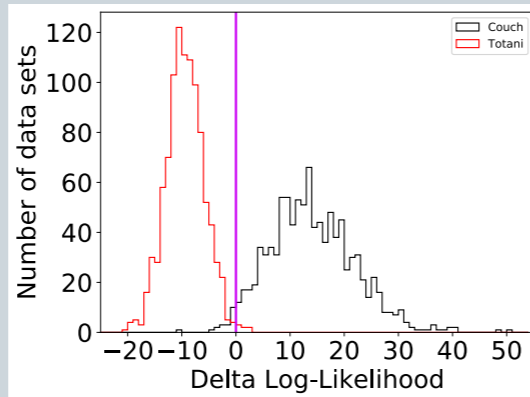
Pairwise Comparison of SN Models

Couch

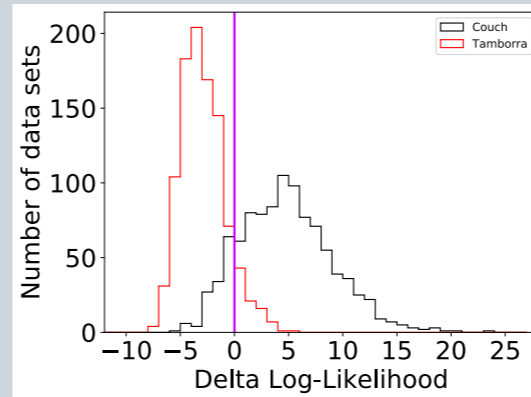
Vartanyan



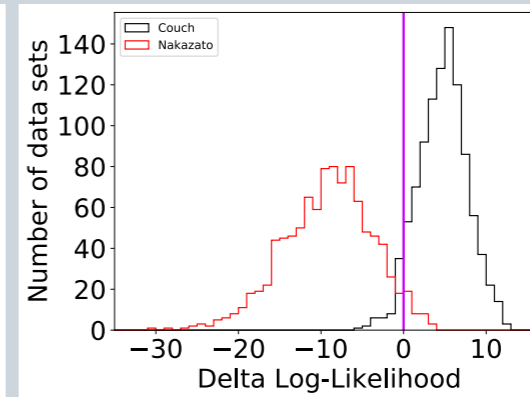
Totani



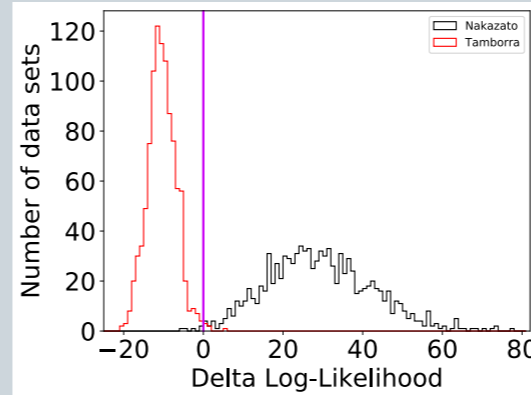
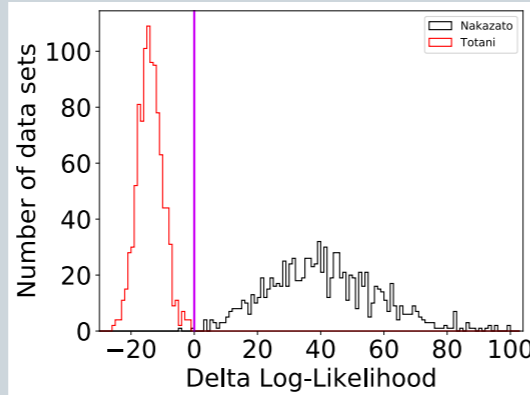
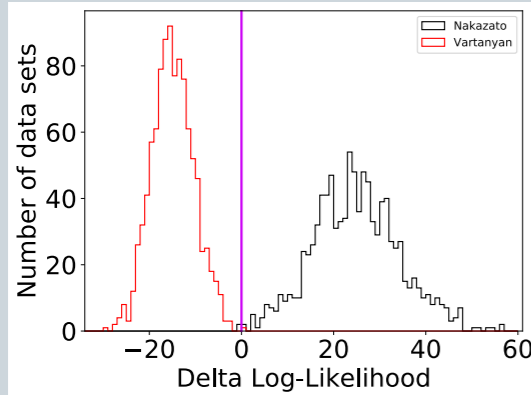
Tamborra



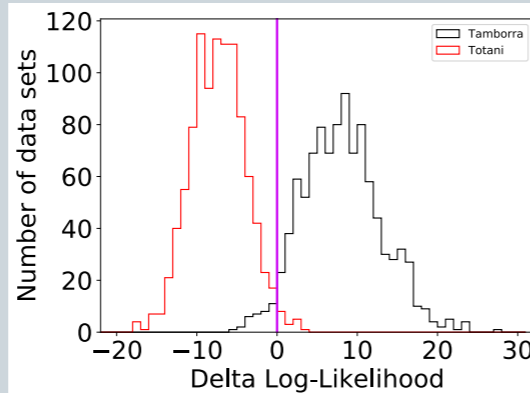
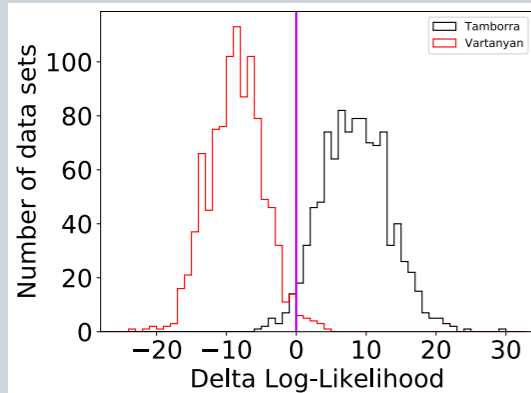
Nakazato



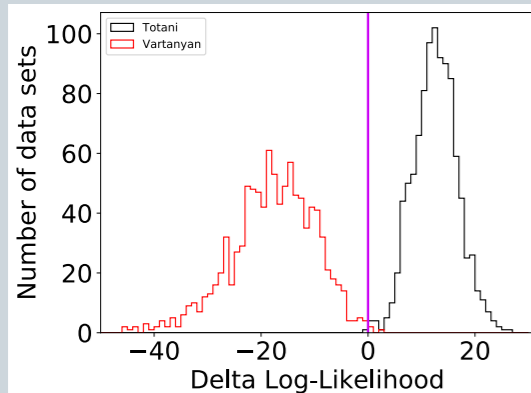
Nakazato



Tamborra



Totani



vertical line indicates $\Delta L = 0$

Normal ordering

$N=100$ events/data set

Model Identification, N=100

		Identified as					
		Normal	Couch	Nakazato	Tamborra	Totani	Vartanyan
True model	Normal						
	Couch	795	57	122	12	14	
	Nakazato	33	961	3	1	2	
	Tamborra	84	0	853	33	30	
	Totani	4	0	16	979	1	
Vartanyan	0	1	17	3	979		

Model Identification, N=100

		Identified as				
		Normal	Couch	Nakazato	Tamborra	Totani
True model	Couch	795	57	122	12	14
	Nakazato	33	961	3	1	2
	Tamborra	84	0	853	33	30
	Totani	4	0	16	979	1
	Vartanyan	0	1	17	3	979

		Identified as				
		Inverted	Couch	Nakazato	Tamborra	Totani
True model	Couch	960	35	4	1	0
	Nakazato	8	992	0	0	0
	Tamborra	0	1	858	21	120
	Totani	3	0	20	977	0
	Vartanyan	0	2	105	1	892

Tamborra model similar to Couch (Vartanyan) model in Normal (Inverted) Ordering.
Other models are separated well!

Model Identification, N=300

		Identified as				
		Normal	Couch	Nakazato	Tamborra	Totani
True model	Couch	982	2	16	0	0
	Nakazato	1	999	0	0	0
	Tamborra	16	0	980	2	2
	Totani	0	0	0	1000	0
	Vartanyan	0	0	0	0	1000

		Identified as				
		Inverted	Couch	Nakazato	Tamborra	Totani
True model	Couch	999	1	0	0	0
	Nakazato	0	1000	0	0	0
	Tamborra	0	0	974	1	25
	Totani	0	0	0	1000	0
	Vartanyan	0	0	8	0	992

Higher statistics reduce random fluctuations & improve accuracy.

... and beyond Hyper-Kamiokande

SN Model Discrimination by Super-K

- Works for any ν detector that can measure t , E of individual events
- For Super-K, scale with detector volume to get a rough estimate
- SK is $8\times$ smaller than HK, so expect similar capability for a SN that is $\sqrt{8} \approx 2.8\times$ closer


Model	Mass		events at 10 kpc*	N=100	N=300
Totani arXiv:astro-ph/9710203	20 M_{\odot}	1D	19716	140 kpc	81 kpc
Nakazato arXiv:1210.6841	20 M_{\odot}	1D	17978	134 kpc	77 kpc
Couch arXiv:1902.01340	20 M_{\odot}	1D	27539	166 kpc	96 kpc
Vartanyan similar to arXiv:1804.00689	9 M_{\odot}	2D	10372	102 kpc	59 kpc
Tamborra arXiv:1406.0006	27 M_{\odot}	3D	25021	158 kpc	91 kpc


*21-34 kpc
for Super-K*

SN Model Discrimination by WATCHMAN

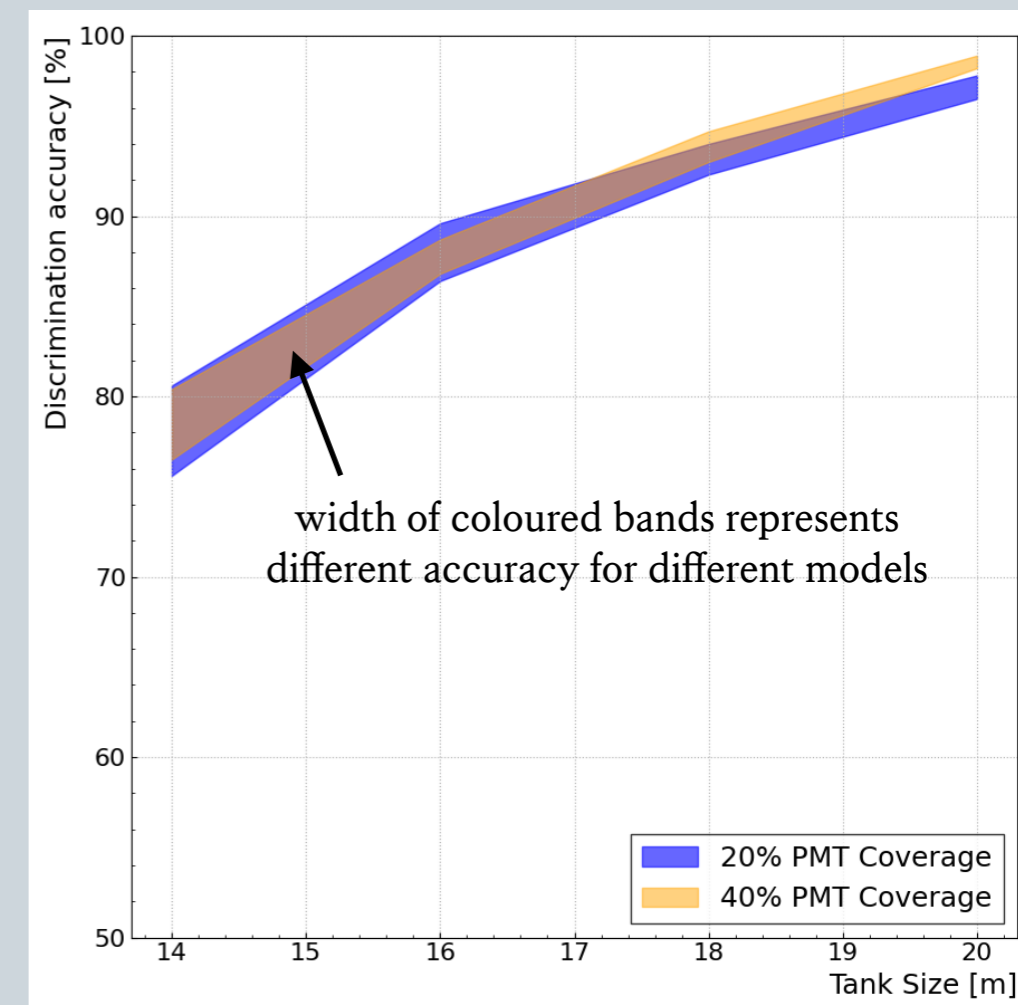
- Works for any ν detector that can measure t , E of individual events
- Detailed analysis by WATCHMAN/NEO finds comparable results (IBD only; using 3 models: Nakazato, Vartanyan, Warren)
- Also used this to benchmark different potential detector configurations: detector volume & photocoverage

PMT Coverage	Tank Size			
	14m	16m	18m	20m
20%	21	39	64	100
40%	21	40	67	104

Number of expected events 

Model discrim. accuracy 

PMT Coverage	Tank Size			
	14m	16m	18m	20m
20%	75.6 – 80.6%	86.4 – 89.6%	92.3 – 94.0%	96.5 – 97.8%
40%	76.5 – 80.4%	86.8 – 88.7%	92.9 – 94.8%	98.2 – 99.0%



Work by Yan-Jie Schnellbach (Liverpool), presented at IOP APP/HEPP/NP Conference 2021

Progenitor Information

- So we can tell apart completely different models ... but what about more closely related ones?
- Necessary to constrain progenitor properties (e.g. mass, metallicity, ...)

Progenitor Information

- So we can tell apart completely different models ... but what about more closely related ones?
- Necessary to constrain progenitor properties (e.g. mass, metallicity, ...)
- Initial study in my PhD thesis:
 - 300 events (60–100 kpc)
 - Nakazato model
 - 13, 20, 30 M_{sol}
 - 20 M_{sol} with 2 metallicities

Progenitor Information

- So we can tell apart completely different models ... but what about more closely related ones?
- Necessary to constrain progenitor properties (e.g. mass, metallicity, ...)
- Initial study in my PhD thesis:
 - 300 events (60–100 kpc)
 - Nakazato model
 - 13, 20, 30 M_{sol}
 - 20 M_{sol} with 2 metallicities

		Identified as			
		13 M_{sol}	20 M_{sol}	20 M_{sol}^*	30 M_{sol}
True model	Normal Mass Ordering				
	13 M_{sol}	878	61	61	0
	20 M_{sol}	17	944	39	0
	20 M_{sol}^*	74	75	850	1
	30 M_{sol}	0	0	0	1000

Progenitor Information

- So we can tell apart completely different models ... but what about more closely related ones?

- Necessary to constrain progenitor properties (e.g. mass, metallicity, ...)

- Initial study in my PhD thesis:

- 300 events (60–100 kpc)
- Nakazato model
- 13, 20, 30 M_{sol}
- 20 M_{sol} with 2 metallicities

		Identified as			
		13 M_{sol}	20 M_{sol}	20 M_{sol}^*	30 M_{sol}
True model	Normal Mass Ordering				
	13 M_{sol}	878	61	61	0
	20 M_{sol}	17	944	39	0
	20 M_{sol}^*	74	75	850	1
30 M_{sol}	0	0	0	1000	

- New pre-print by Olsen & Qian ([arXiv:2202.09975](https://arxiv.org/abs/2202.09975))

- Compare simulations with different progenitor masses (9.6, 18.6, 20, 27 M_{sol}) and equations of state (SFHo, LS220)
- Simpler detector model (Super-K-like), more sophisticated statistical approach
- Maximum distance to distinguish models is 10–25 kpc

Beyond Neutrinos

- ♦ Gravitational waves
 - ♦ Complementary ways of studying the SN core!
 - ♦ A few simulations provide both ν and GW signals
(Warren et al., [DOI:10.3847/1538-4357/ab97b7](https://doi.org/10.3847/1538-4357/ab97b7)) → enables cross checks and/or joint likelihood function for improved accuracy
- ♦ Electromagnetic
 - ♦ Complementary ways of extracting information on the progenitor!
 - ♦ ... and its core? (Barker et al., [arXiv:2102.01118](https://arxiv.org/abs/2102.01118))
 - ♦ Relies on stellar evolution code?
- ♦ Plenty of room for discussion today!

Conclusions

- ♦ Hyper-Kamiokande (high event rate & event-by-event energy information) can tell SN models apart even at ~100 kpc distance
 - ♦ $\geq 80\%$ able to identify true model with 100 evts
 - ♦ $\geq 97\%$ accuracy with 300 evts
 - ♦ If mis-ID: only by very narrow margin \rightarrow can at least narrow down list of possible models
- ♦ Similar techniques can be adopted by other neutrino detectors (e.g. WATCHMAN, SK-like)
- ♦ First analyses show potential to identify progenitor properties
- ♦ Complementarity with GW & EM possible