

Bayesian Analysis with KIDS Energy Density Functional

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in collaboration with BUD :

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1st detection of GW from a BNS merger

GW170817

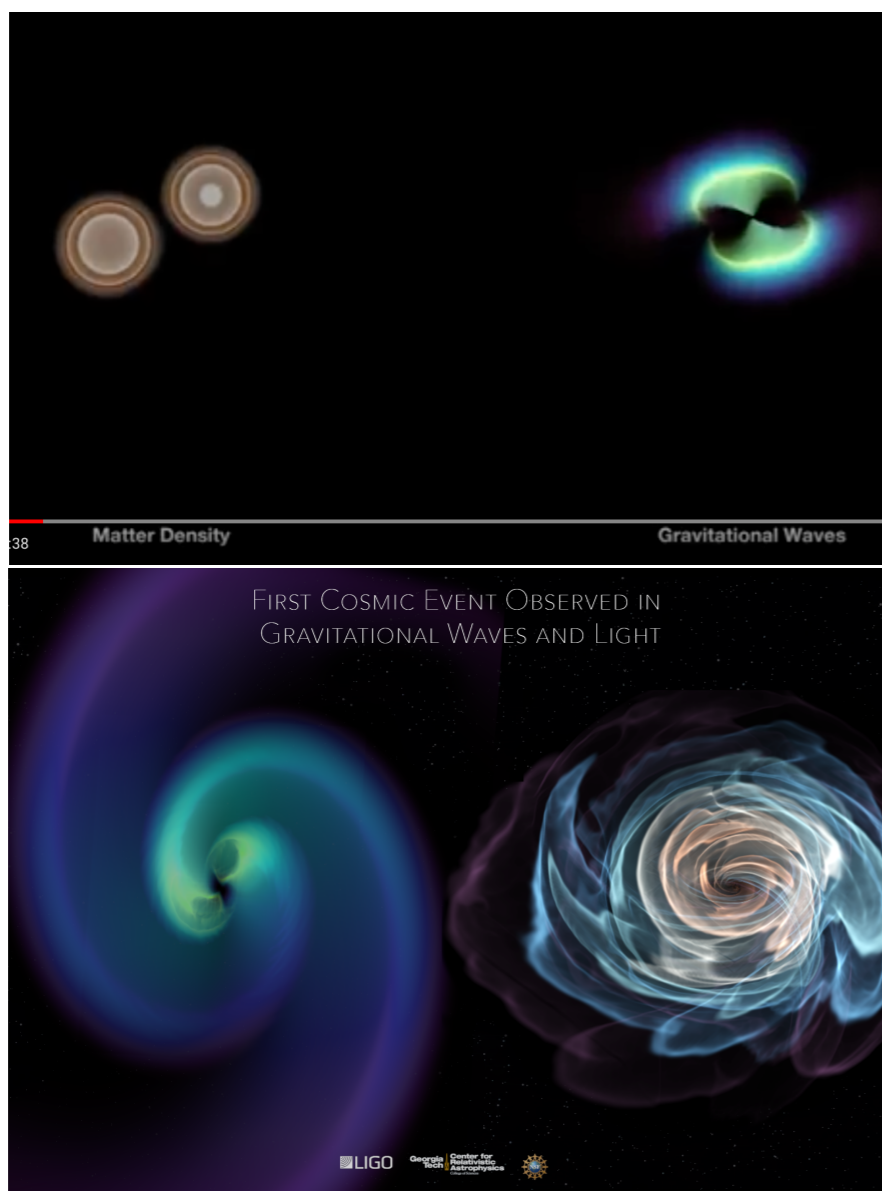
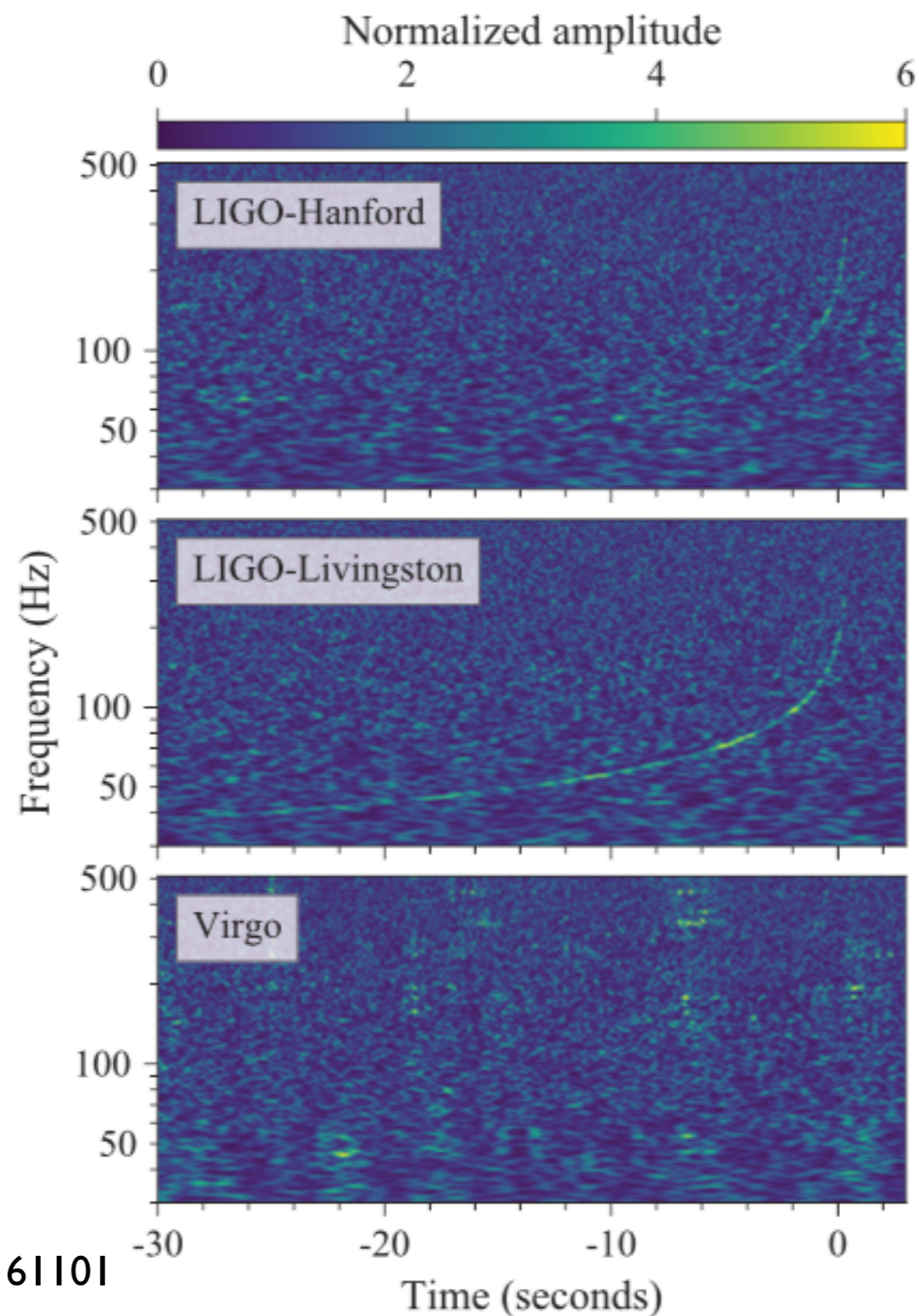


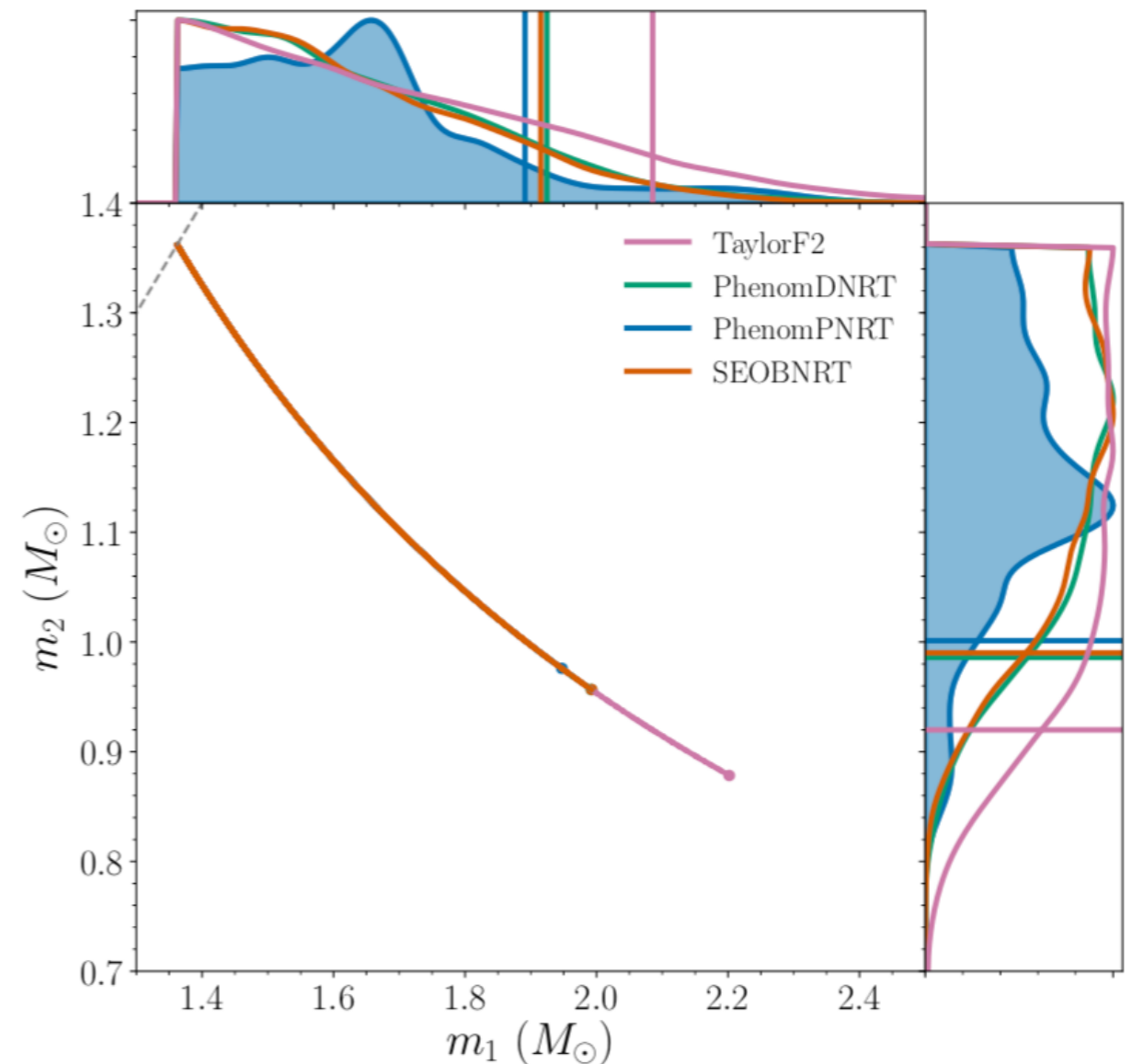
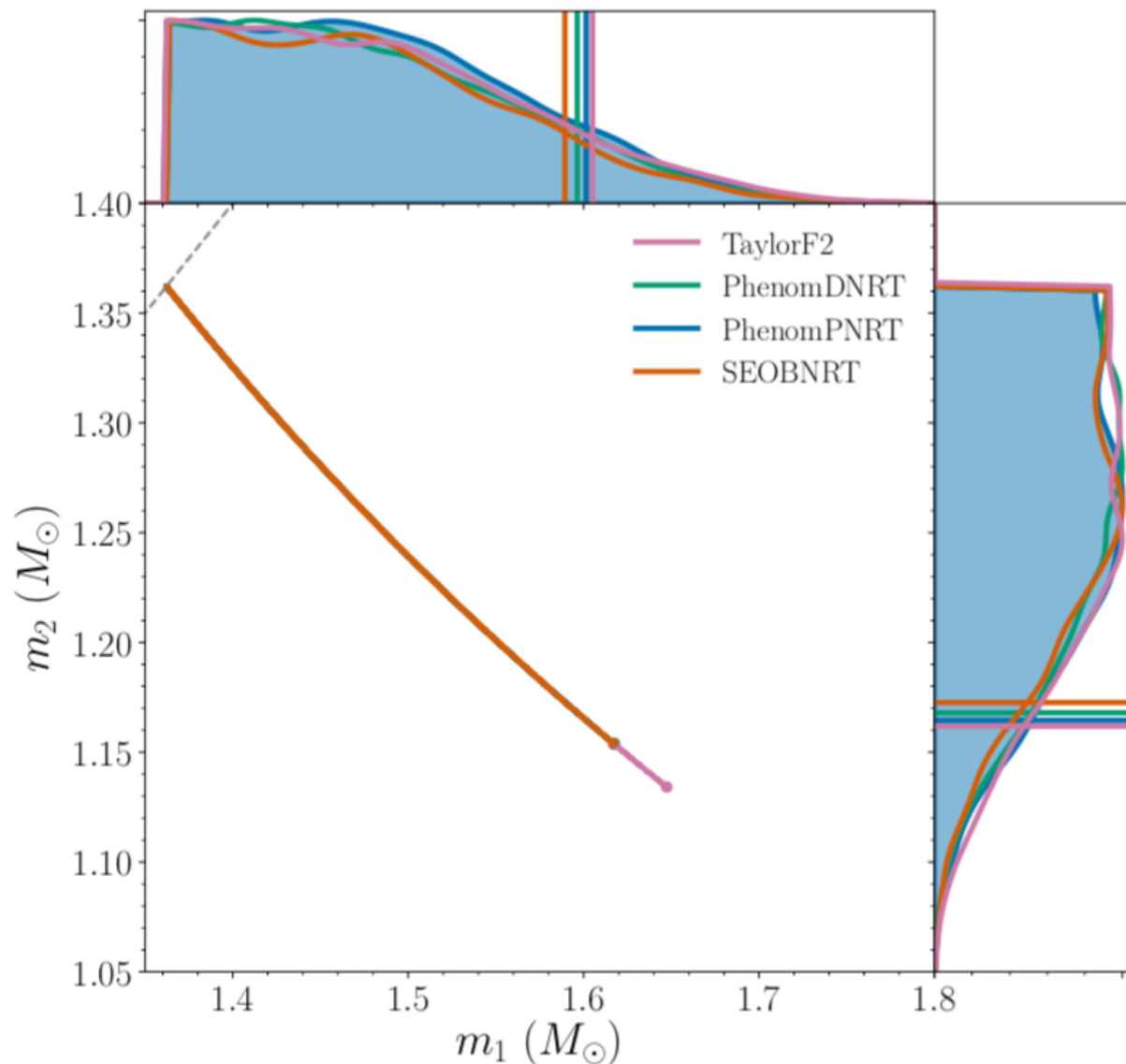
Image credit: Karan Jani/Georgia Tech.



Bayesian Analysis on GW observation (I)

low spin prior : $\chi \leq 0.05$
 $m_1 = 1.36 \sim 1.60 M_\odot$
 $m_2 = 1.16 \sim 1.36 M_\odot$

high spin prior : $\chi \leq 0.89$
 $m_1 = 1.36 \sim 1.89 M_\odot$
 $m_2 = 1.00 \sim 1.36 M_\odot$

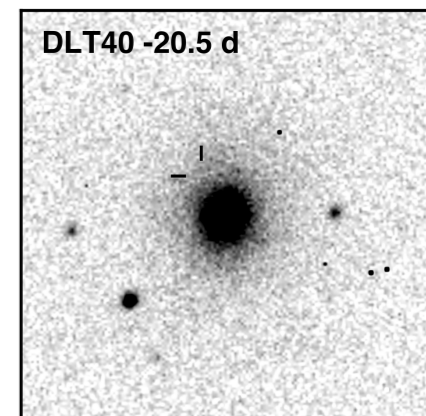
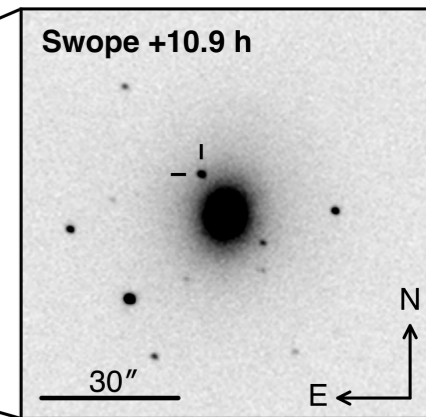
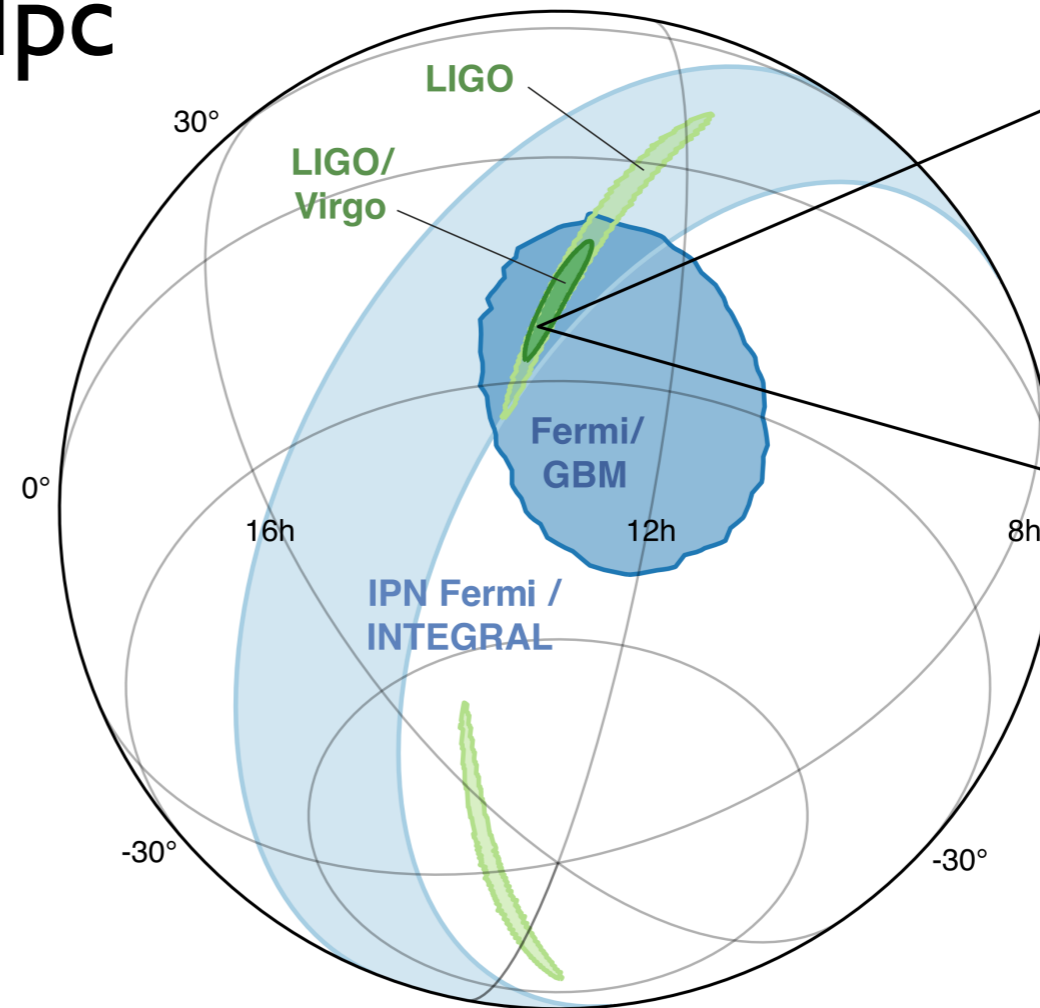


Abbott et al. (LSC and Virgo), arxiv:1805.11579 (PhysRevX.9.011001)

Bayesian Analysis on GW observation (2)

$D_L \sim 40 \text{ Mpc}$

ApJL.848.L12(2017)

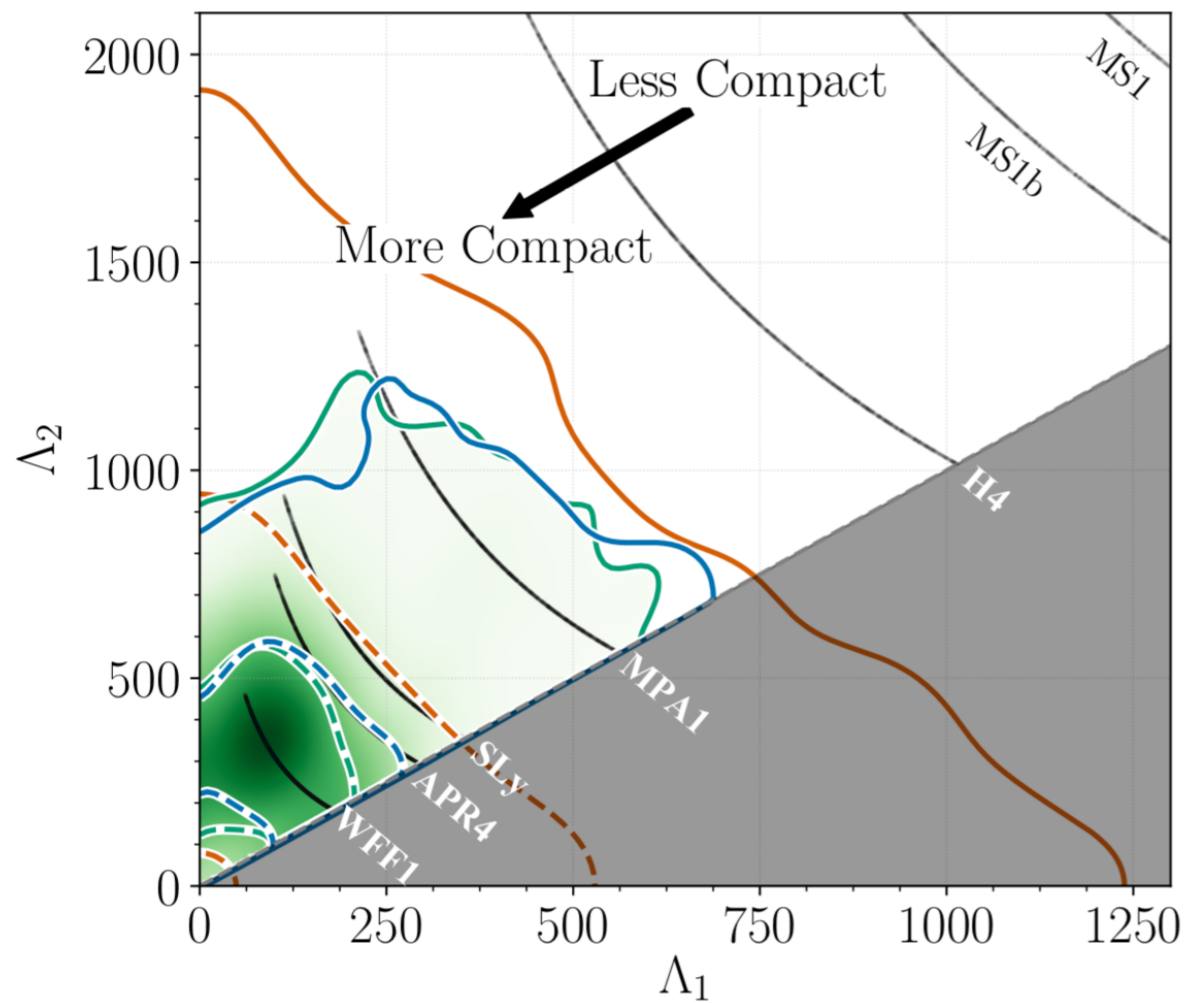


AT2017gfo in NGC4993



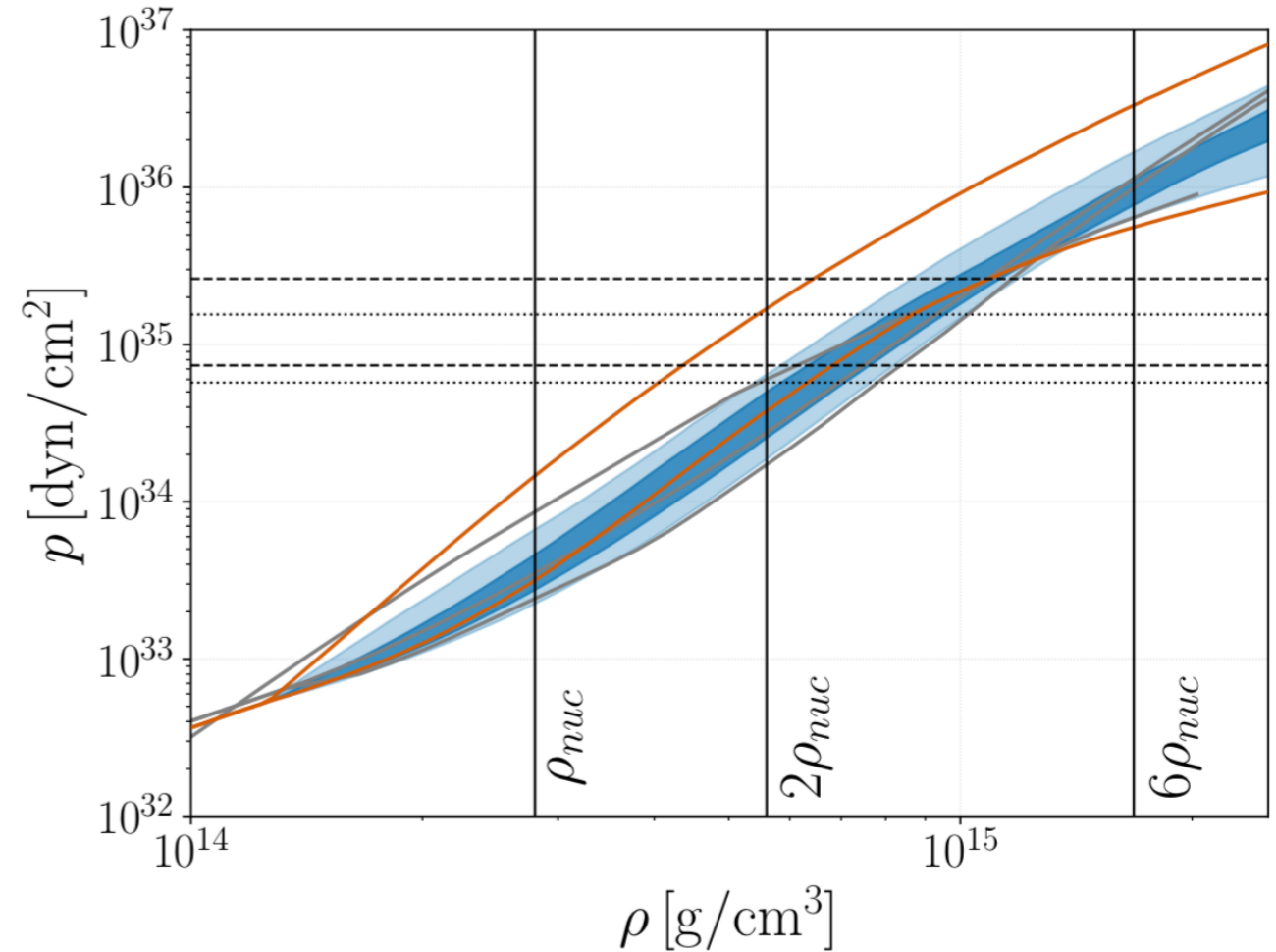
Bayesian Analysis on GW observation (3)

$$\Lambda(1.4M_{\odot}) = 190^{+390}_{-120}$$



$$P(2 \rho_{\text{nuc}}) = 3.5^{+2.7}_{-1.7} \times 10^{34} \text{ dyne/cm}^2$$

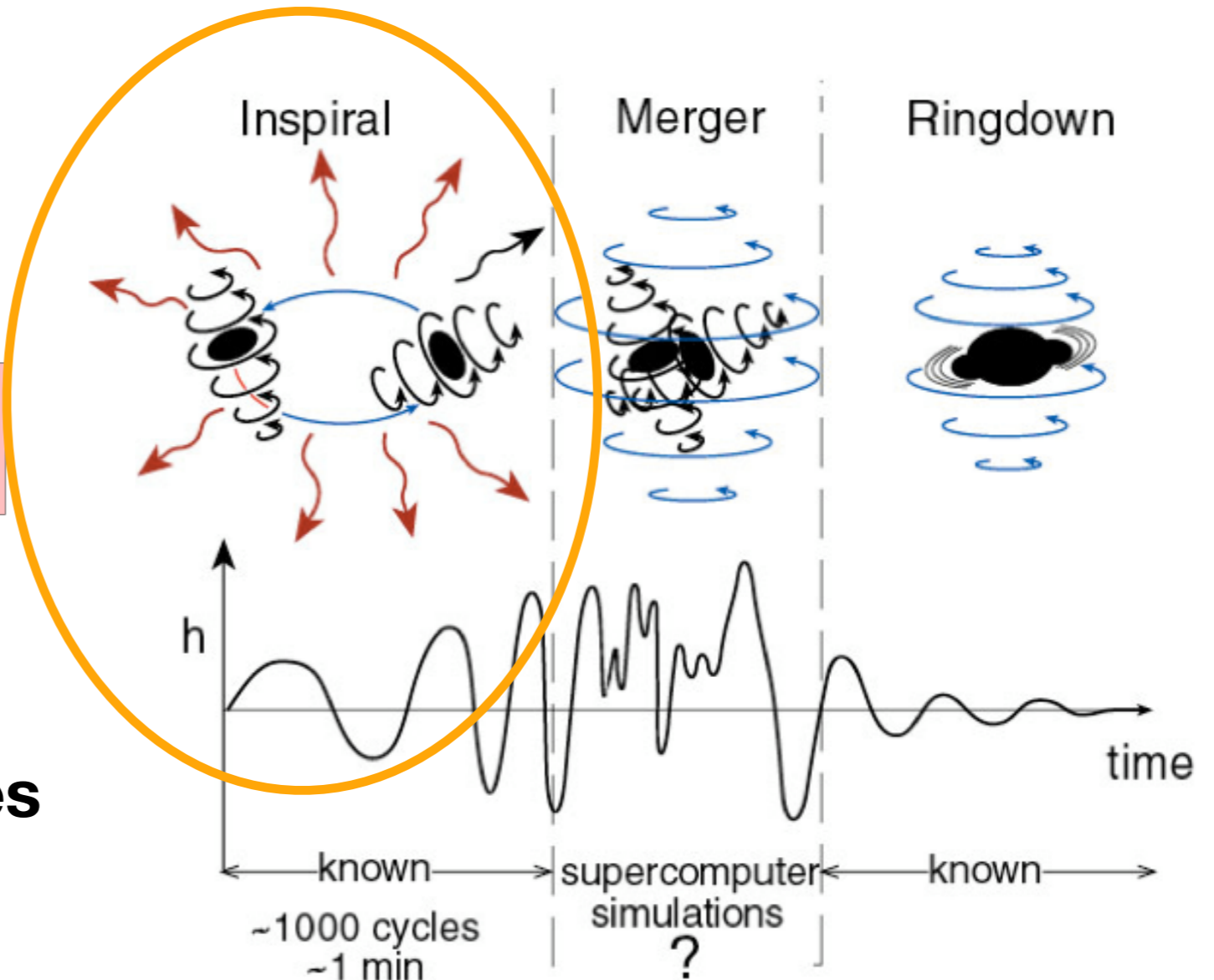
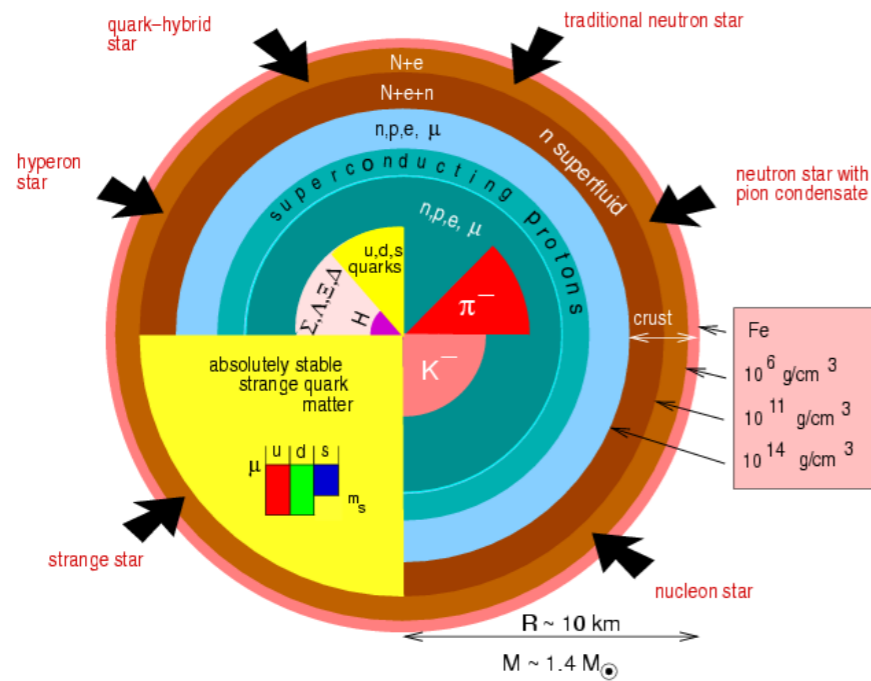
$$P(6 \rho_{\text{nuc}}) = 9.0^{+7.9}_{-2.6} \times 10^{35} \text{ dyne/cm}^2$$



Abbott et al. (LSC and Virgo), arxiv:1805.11581 (PhysRevLett.121.161101)

$$\rho_{\text{nuc}} = 2.8 \times 10^{14} \text{ g/cm}^3$$

NS Tidal deformability



perturbative approaches

$$Q_{ij} = -\lambda \mathcal{E}_{ij}$$

$$\lambda = \frac{2}{3} \frac{R^5}{G} k_2$$

λ : Tidal deformability
 k_2 : Tidal Love number

KIDS Energy Density Functional

KIDS Energy density functional form

Phys. Rev. C 97, 014312 (2018)

$$\mathcal{E}(\rho, \delta) = \frac{E(\rho, \delta)}{A} = \mathcal{T}(\rho, \delta) + \sum_{i=0}^{N-1} c_i(\delta) \rho^{1+i/3} \quad \delta = \frac{\rho_n - \rho_p}{\rho}$$

$$\mathcal{T}(\rho, \delta) = \frac{3}{5} \left[\frac{\hbar^2}{2m_p} \left(\frac{1-\delta}{2} \right)^{5/3} + \frac{\hbar^2}{2m_n} \left(\frac{1+\delta}{2} \right)^{5/3} \right] (3\pi^2 \rho)^{2/3}$$

$c_i(\delta) = \alpha_i + \beta_i \delta^2$ to be determined by fitting to the observables

at zero temperature $k_F = (3\pi^2 \rho/2)^{1/3}$ $k_{F_\tau} = k_F (1 + \tau\delta)^{1/3}$

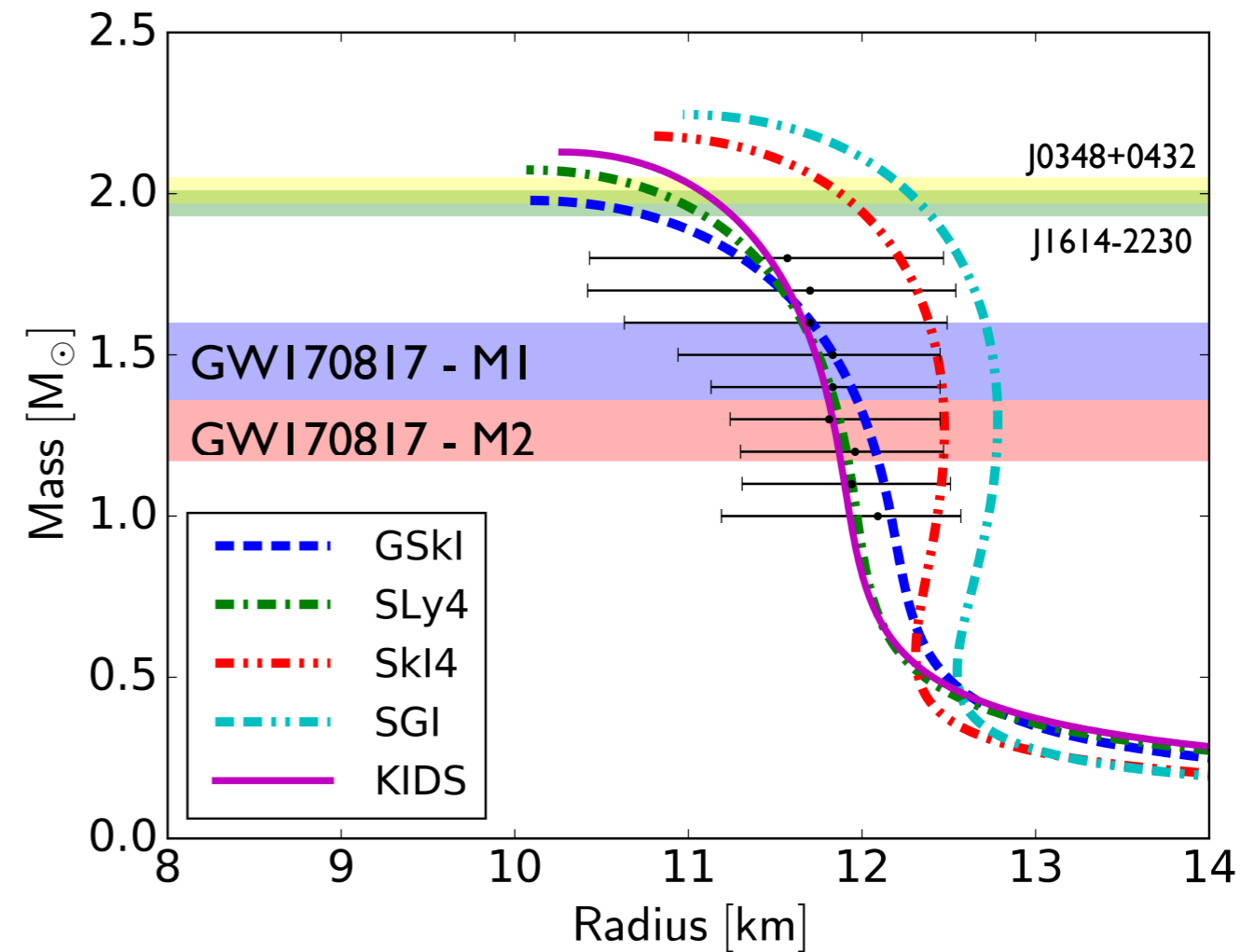
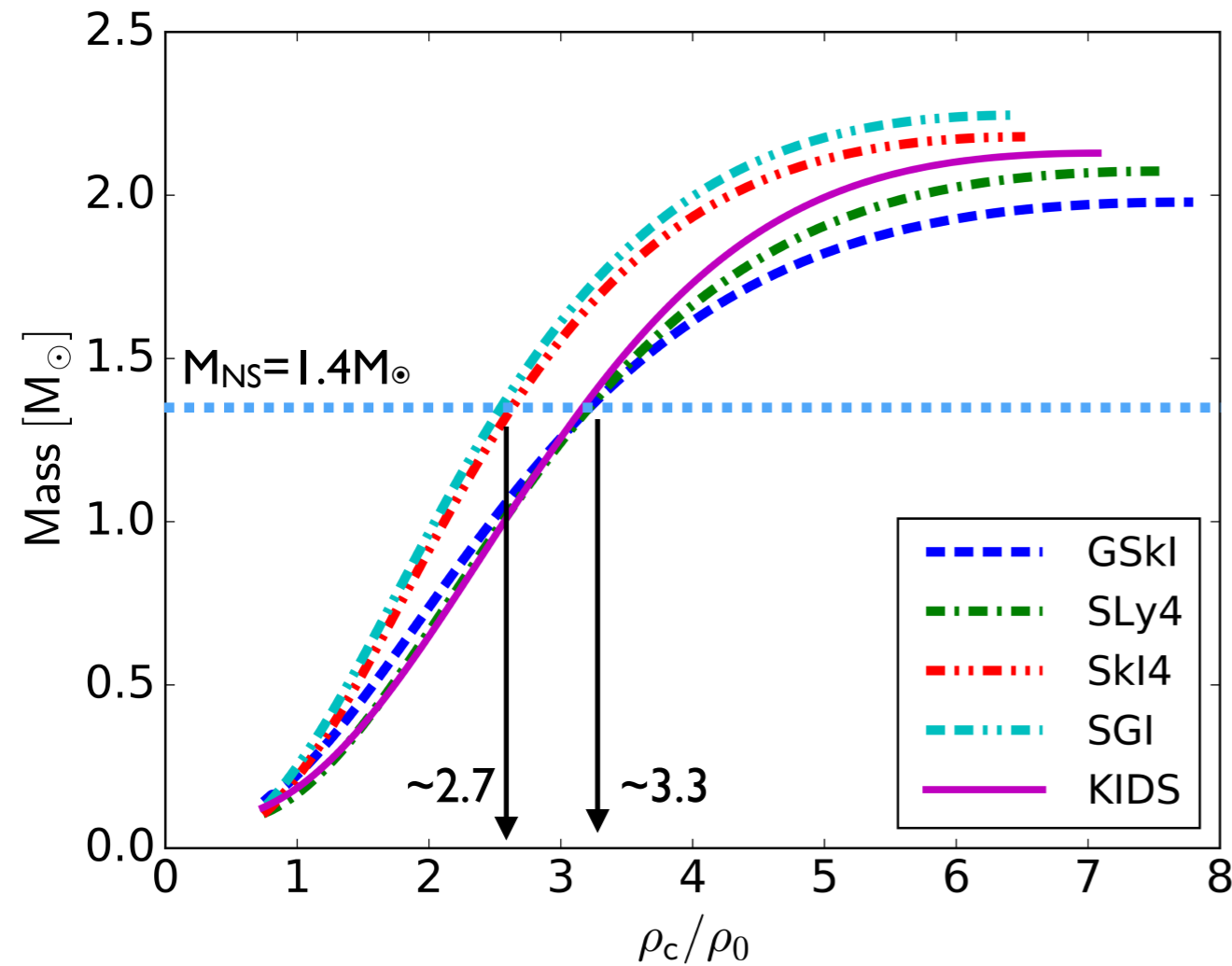
Constraints on Nuclear EoS

- Nuclear data: hundreds of models (Skyrme force, RMF, ...)
- Neutron star maximum mass
 $1.97 \pm 0.04 M_{\odot}$ [Nature 467, 1081 (2010)]
 $2.01 \pm 0.04 M_{\odot}$ [Science 340, 448 (2013)]
- II experimental/empirical data for nuclear matter around saturation density [Phys.Rev. C 85, 035201 (2012)]

Constraint	Quantity	Eq.	Density Region	Range of constraint exp/emp	Range of constraint from CSkP	Ref.
SM1	K_{\circ}	(7), (15)	ρ_{\circ} (fm^{-3})	200 – 260 MeV	202.0 – 240.3 MeV	[64]
SM2	$K' = -Q_{\circ}$	(8), (16)	ρ_{\circ} (fm^{-3})	200 – 1200 MeV	362.5 – 425.6 MeV	[65]
SM3	$P(\rho)$	(6)	$2 < \frac{\rho}{\rho_{\circ}} < 3$	Band Region	see Fig. 1	[78]
SM4	$P(\rho)$	(6)	$1.2 < \frac{\rho}{\rho_{\circ}} < 2.2$	Band Region	see Fig. 2	[80]
PNM1	$\frac{E_{PNM}}{E_{PNM}^{\circ}}$	(31)	$0.014 < \frac{\rho}{\rho_{\circ}} < 0.106$	Band Region	see Fig. 3	[39, 40]
PNM2	$P(\rho)$	(6)	$2 < \frac{\rho}{\rho_{\circ}} < 3$	Band Region	see Fig. 5	[78]
MIX1	J	(9)	ρ_{\circ} (fm^{-3})	30 – 35 MeV	30.0 – 35.5 MeV	[44]
MIX2	L	(10)	ρ_{\circ} (fm^{-3})	40 – 76 MeV	48.6 – 67.1 MeV	[101]
MIX3	$K_{\tau, \nu}$	(21)	ρ_{\circ} (fm^{-3})	-760 – -372 MeV	-407.1 – -360.1 MeV	[107]
MIX4	$\frac{S(\rho_{\circ}/2)}{J}$	-	ρ_{\circ} (fm^{-3})	0.57 – 0.86	0.61 – 0.67	[110]
MIX5	$\frac{3P_{PNM}}{L\rho_{\circ}}$	(41)	ρ_{\circ} (fm^{-3})	0.90 – 1.10	1.02 – 1.10	[112]

Mass-Radius relations

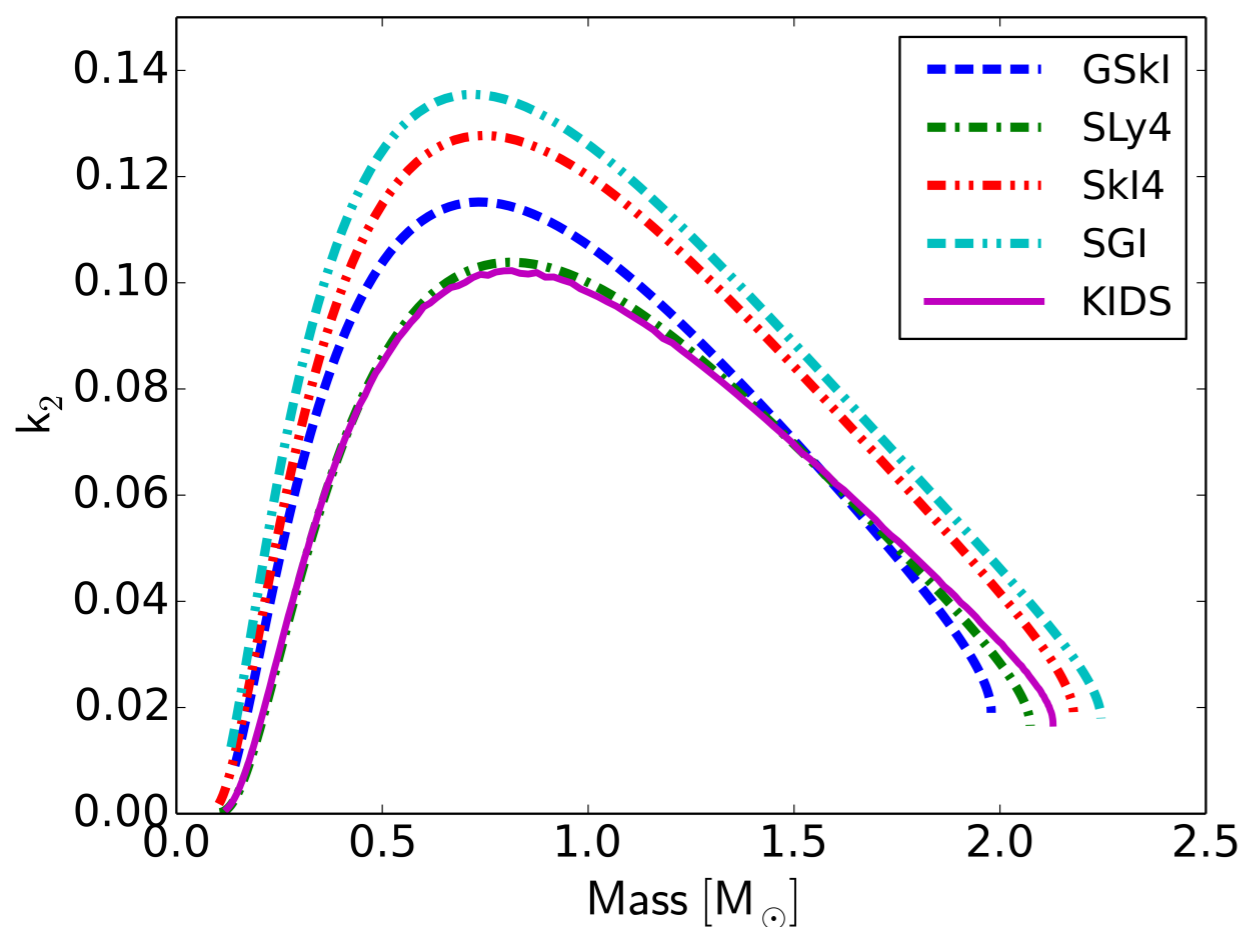
Kim et al., PhysRevC.98.065805 (2018)



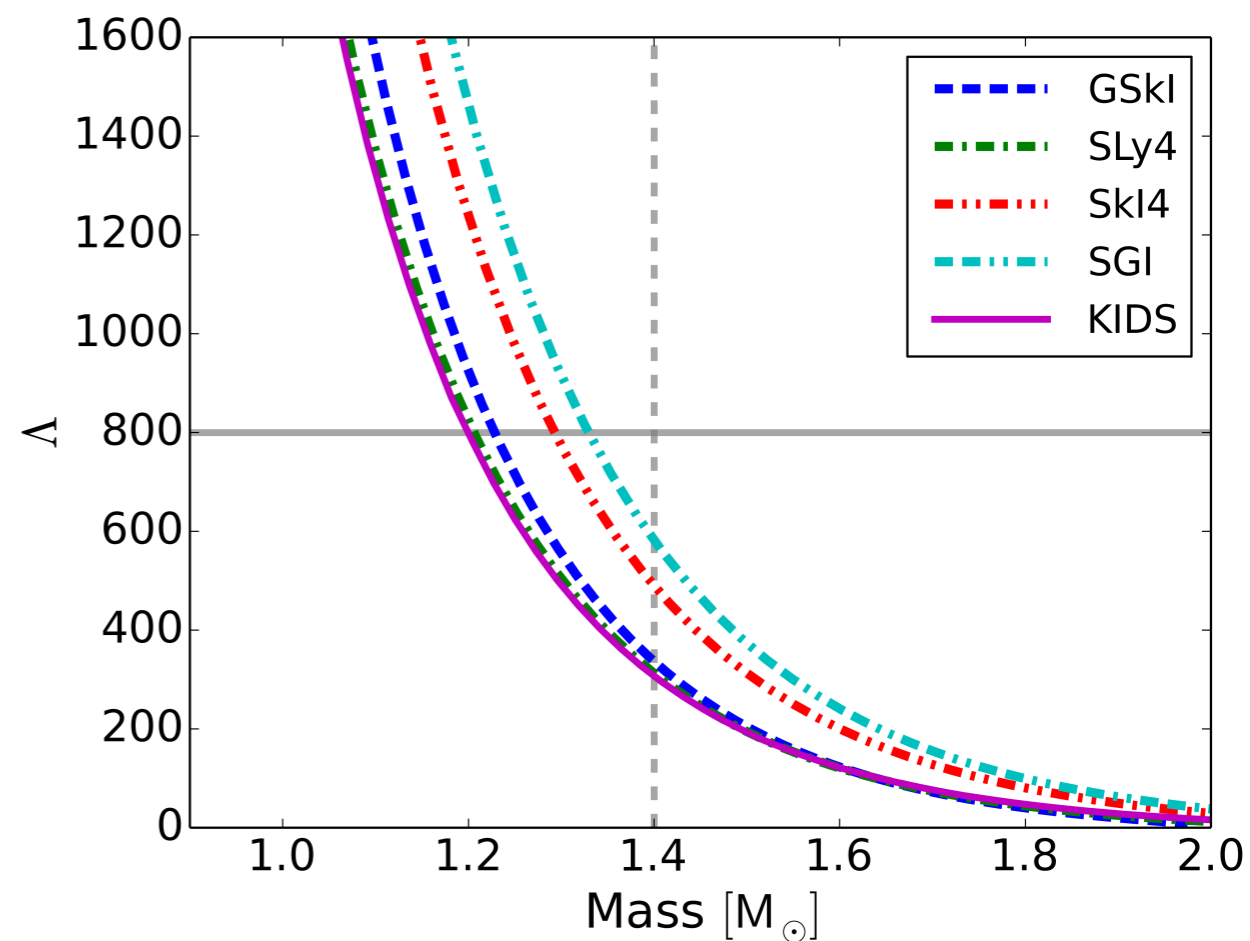
GW170817 - Abbott et al. (LSC and Virgo), arxiv:1805.11579

- $M_{\text{chirp}} = 1.188 M_{\odot}$
- low spin prior : $M_1 = 1.36 \sim 1.60 M_{\odot}$, $M_2 = 1.16 \sim 1.36 M_{\odot}$
- high spin prior : $M_1 = 1.36 \sim 1.89 M_{\odot}$, $M_2 = 1.00 \sim 1.36 M_{\odot}$

Tidal deformability of a NS



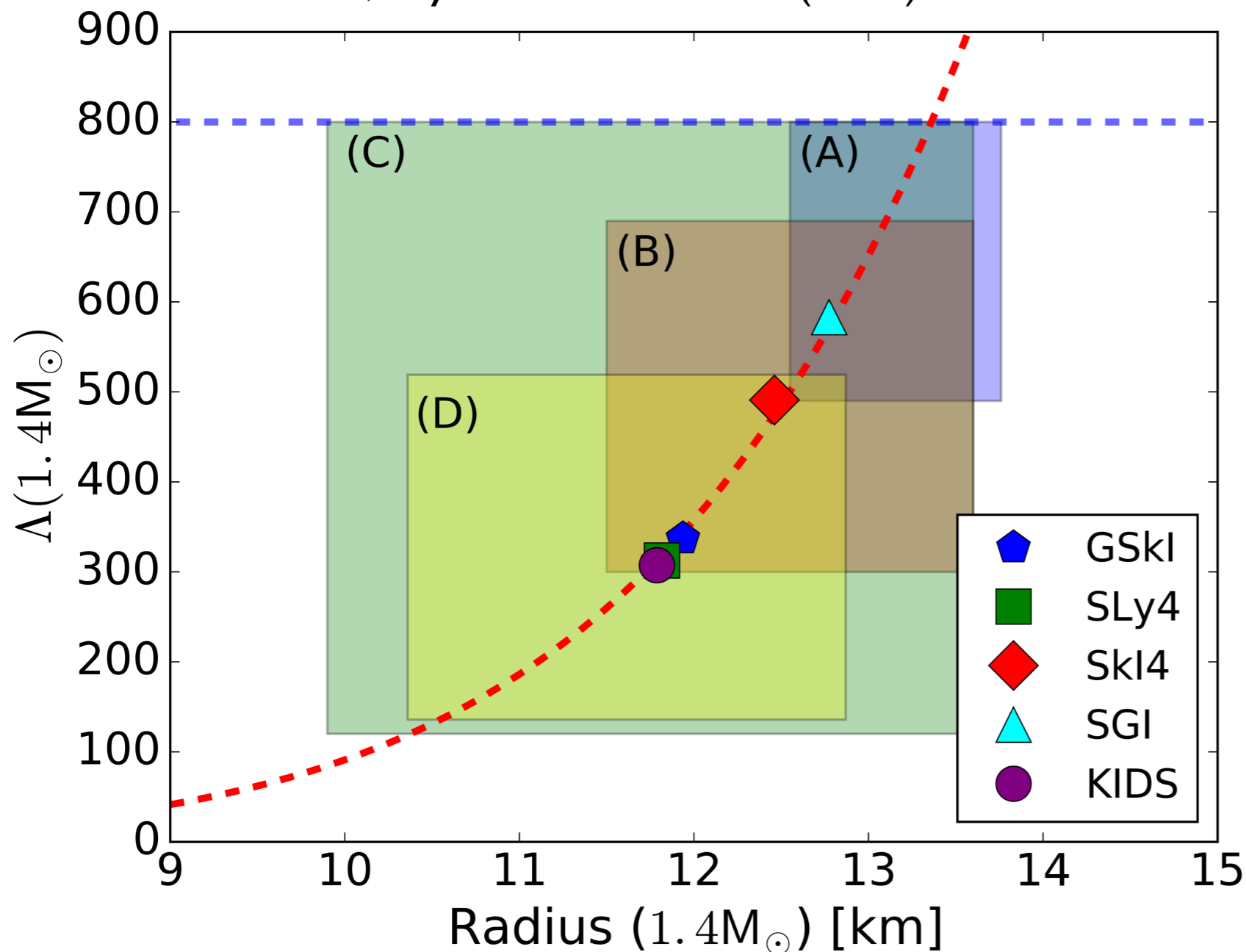
Kim et al., PhysRevC.98.065805 (2018)



GW170817 - Abbott et al. (LSC and Virgo), arxiv:1805.11579
- Mchirp = 1.188 M_\odot
- low spin prior : $\Lambda = 300^{+500}_{-190}$ (symmetric) / 300^{+420}_{-230} (HPD)
- high spin prior : $\Lambda = 0 \sim 630$

Tidal Deformability of 1.4 Msun

Kim et al., PhysRevC.98.065805 (2018)

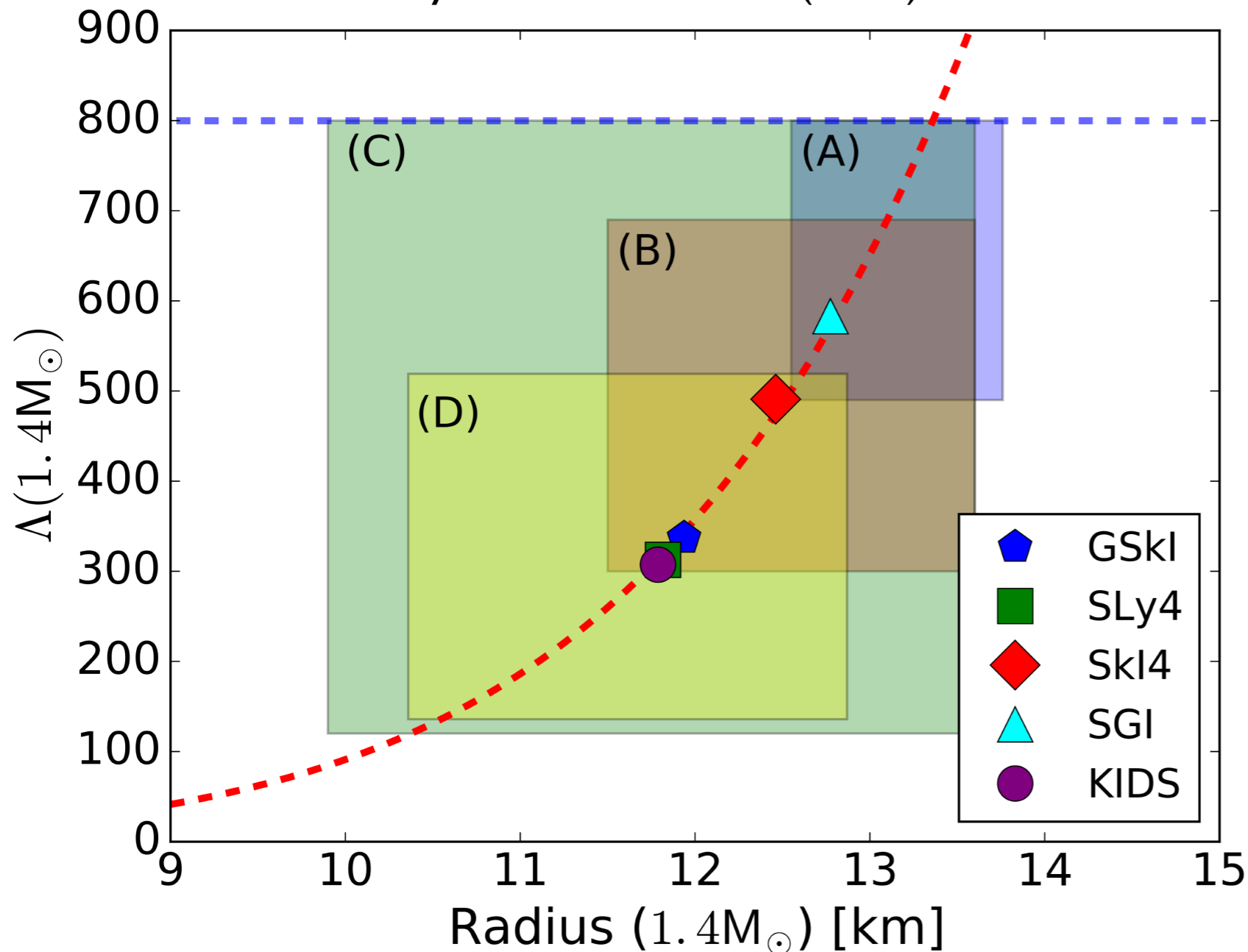


- [A] F.J. Fattoyev, J. Piekarewicz, and C.J. Horowitz, PRL.120.172702 (arXiv:1711.06615v2)
 [B] P.G. Krastev, and B.-A. Li, arXiv:1801.04620v1
 [C] E. Annala, T. Gorda, A. Kurkela, and A. Vuorinen, PRL.120.172703 (arXiv:1711.02644v2)
 [D] Y. Lim and J. Holt, PRL.121.062701 (arXiv:1803.02803v2)

Red line: $\Lambda(1.4M_{\odot}) = 2.88 * 10^{-6} (R/\text{km})^{7.5}$ (fitting function in [C])

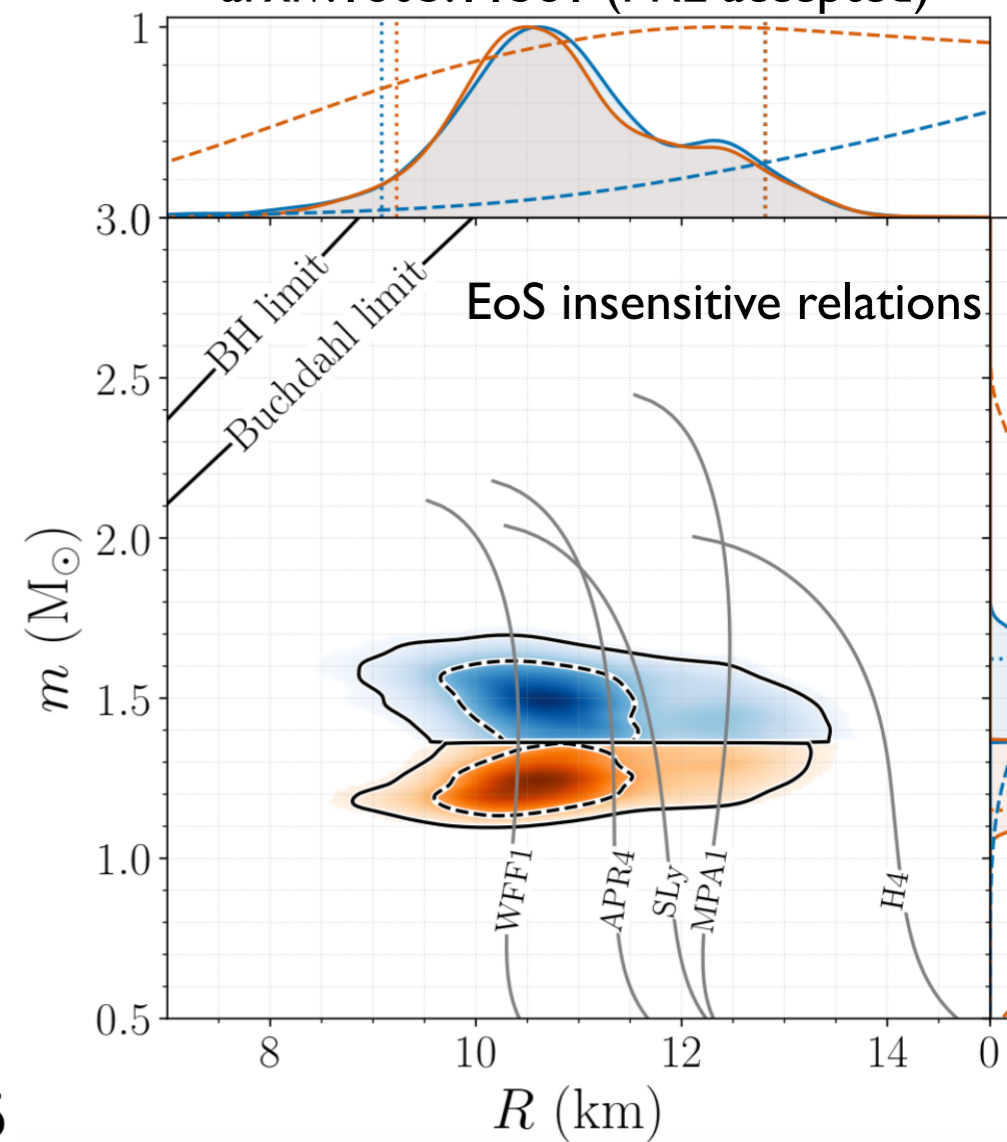
Tidal Deformability of 1.4 Msun

Kim et al., PhysRevC.98.065805 (2018)



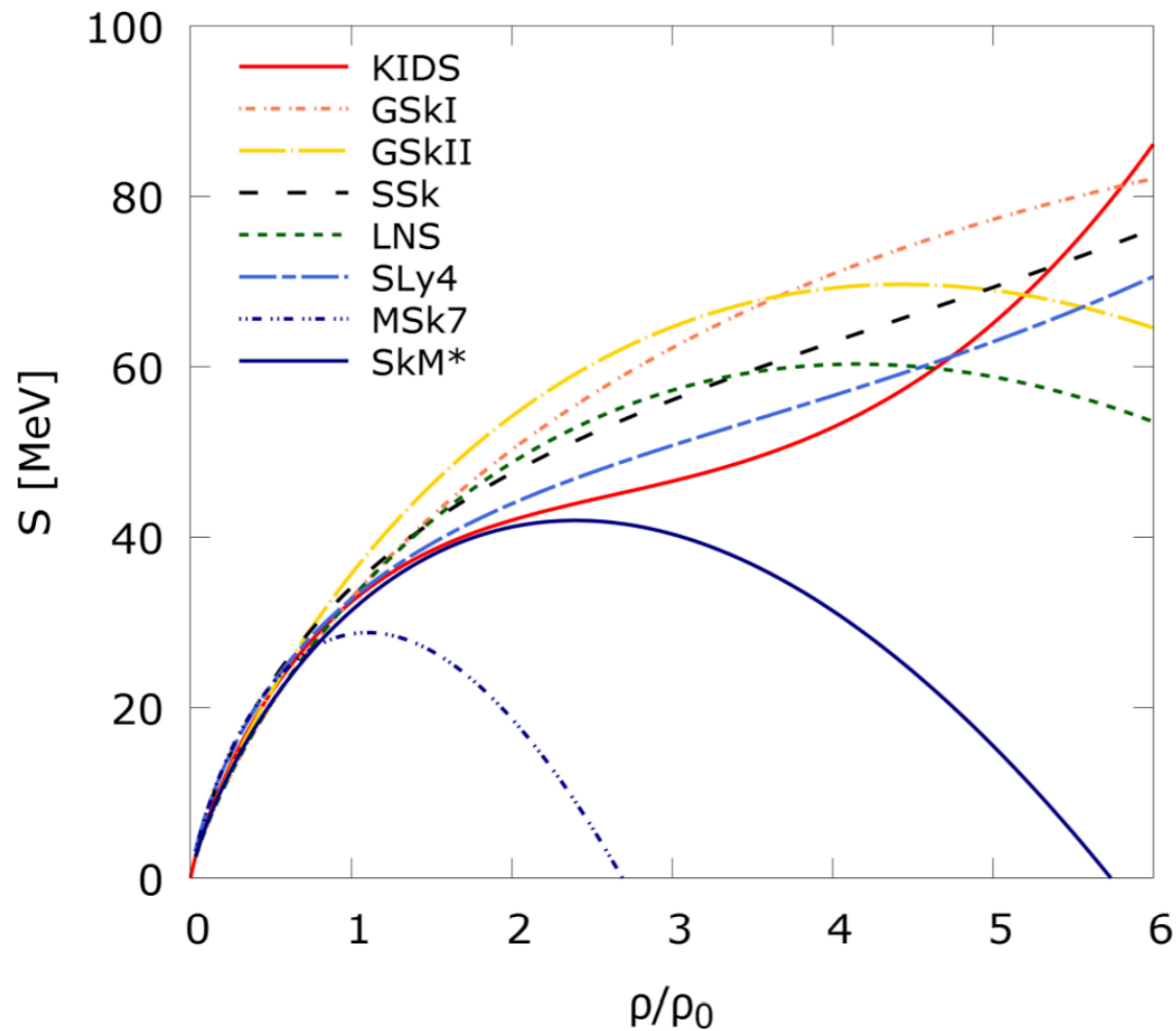
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Abbott et al. (LSC and Virgo),
arxiv:1805.11581 (PRL accepted)

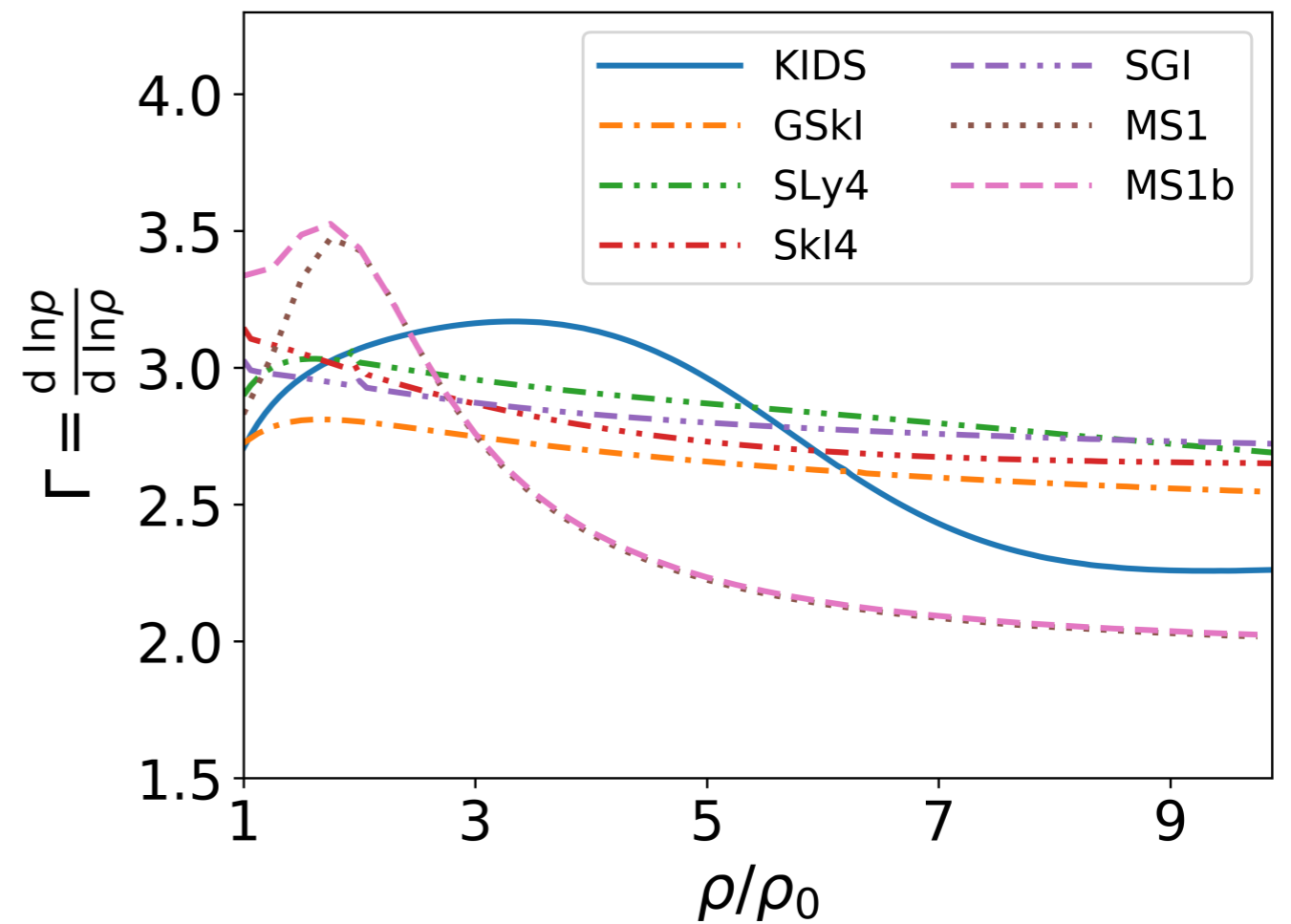


Nuclear Properties

H. Gil, et al., PRC, 99,064319 (2019)



Y.-M. Kim et al., in preparation

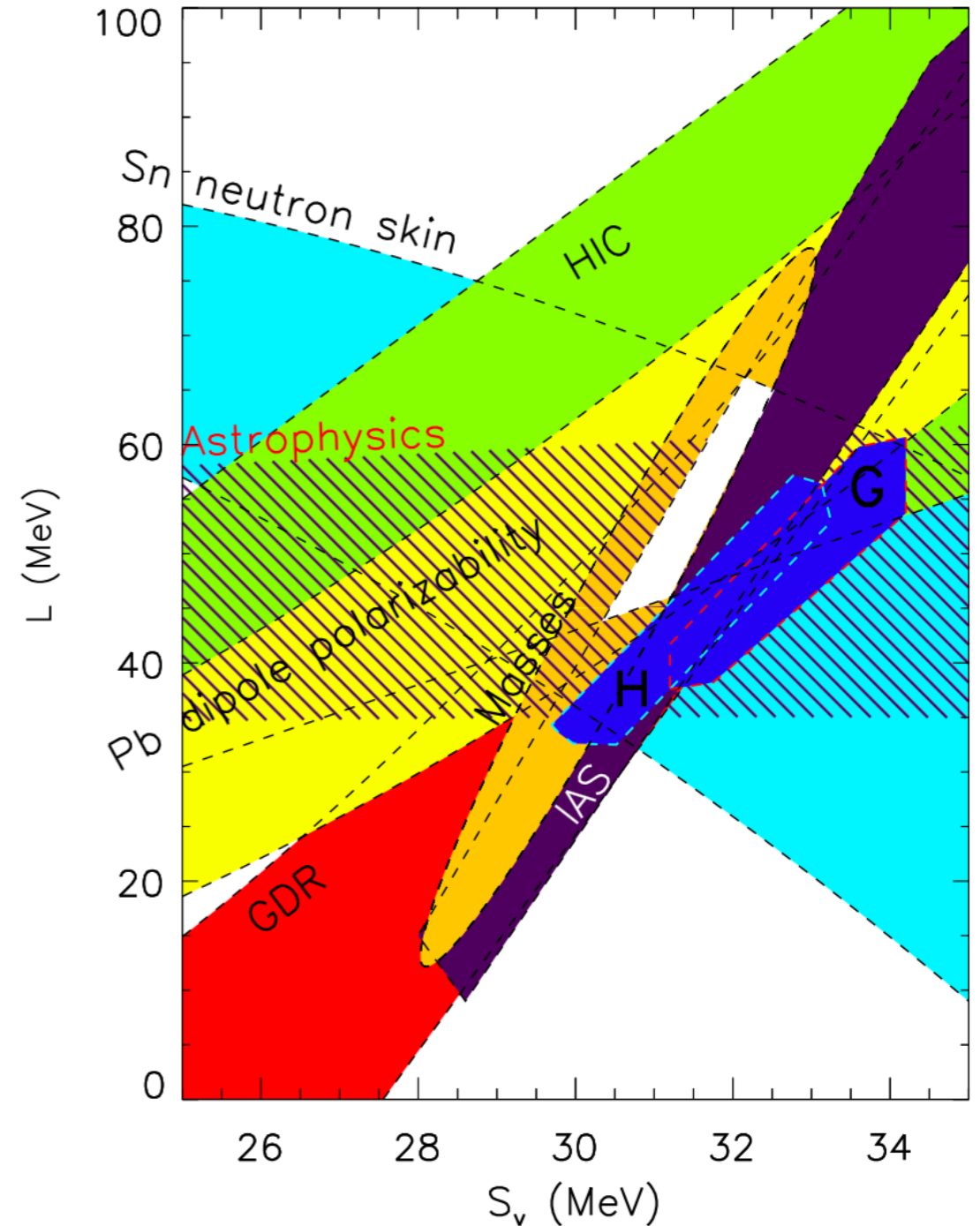
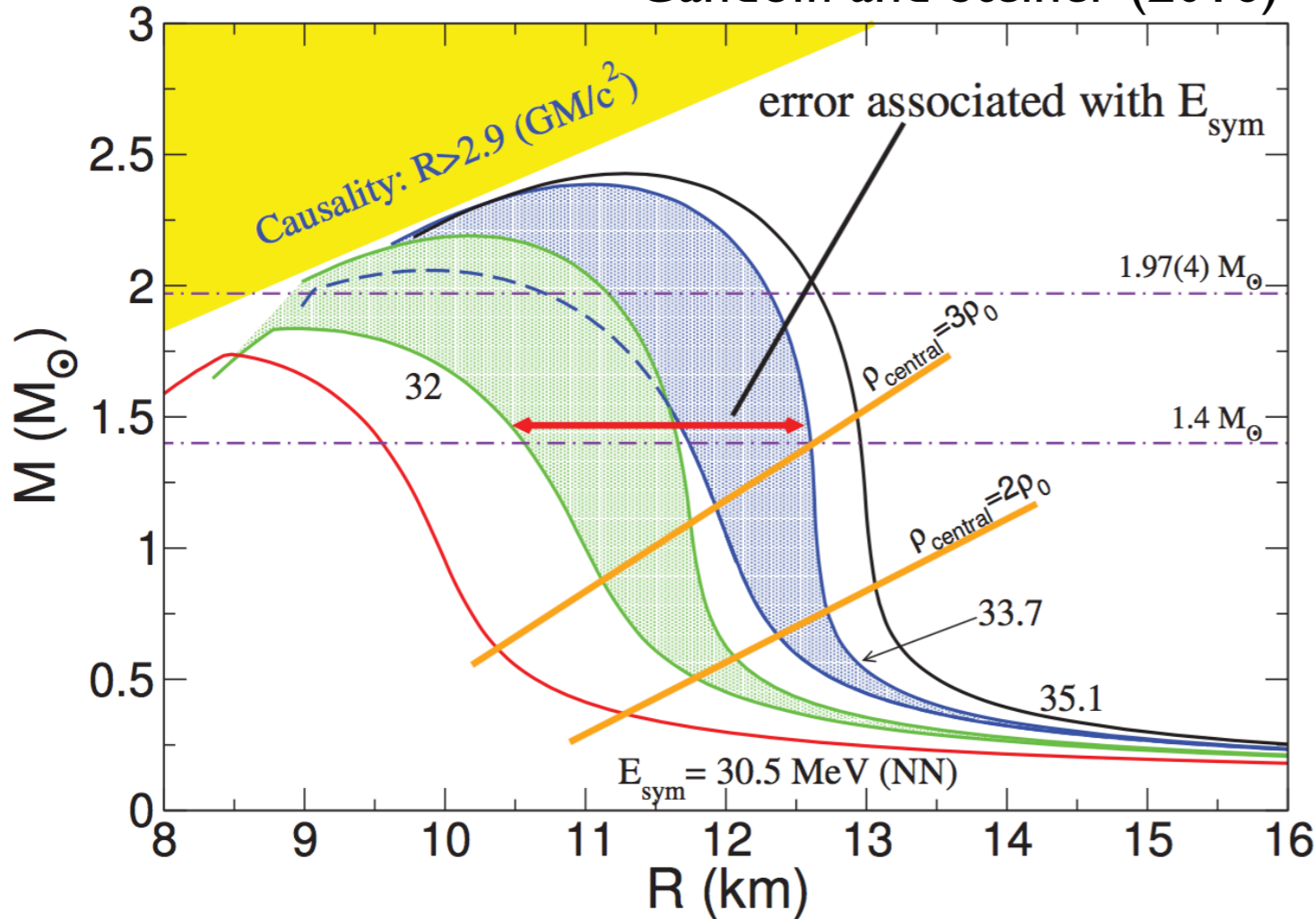


$$\mathcal{S}(\rho) = \left. \frac{1}{2} \frac{\partial^2}{\partial \delta^2} \mathcal{E}(\rho, \delta) \right|_{\delta=0} = J + Lx + \frac{1}{2} K_{\text{sym}} x^2 + \frac{1}{6} Q_{\text{sym}} x^3 + \frac{1}{24} R_{\text{sym}} x^4 + O(x^5), \quad \text{where } x = (\rho - \rho_0)/(3\rho_0).$$

Nuclear Symmetry Energy

J.M. Lattimer / Nuclear Physics A 928 (2014) 276–295

Gandolfi and Steiner (2016)



$$\frac{E}{A} = E(\rho, \delta = 0) + E_{sym}(\rho)\delta^2 + \mathcal{O}(\delta^4) + \dots,$$

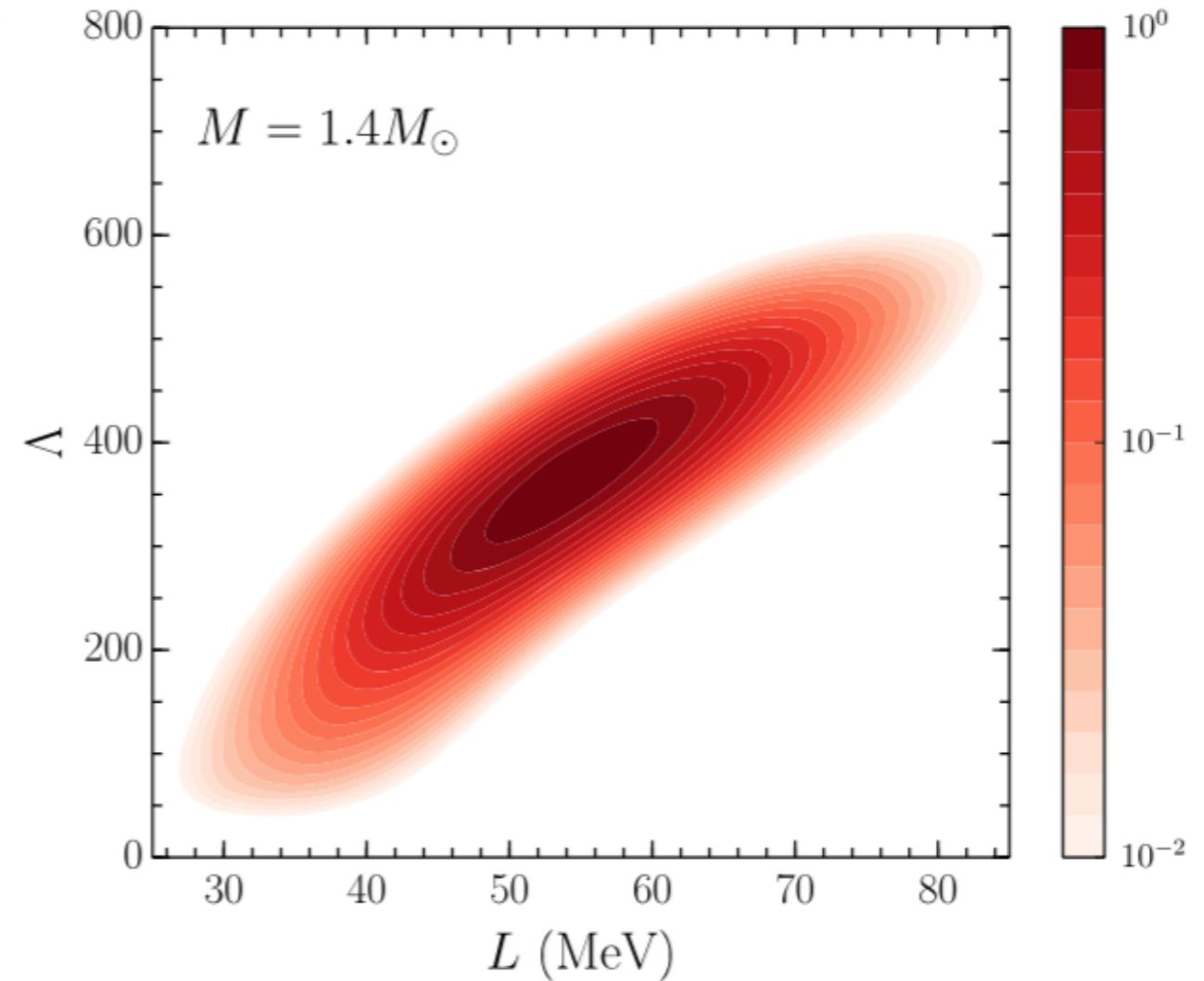
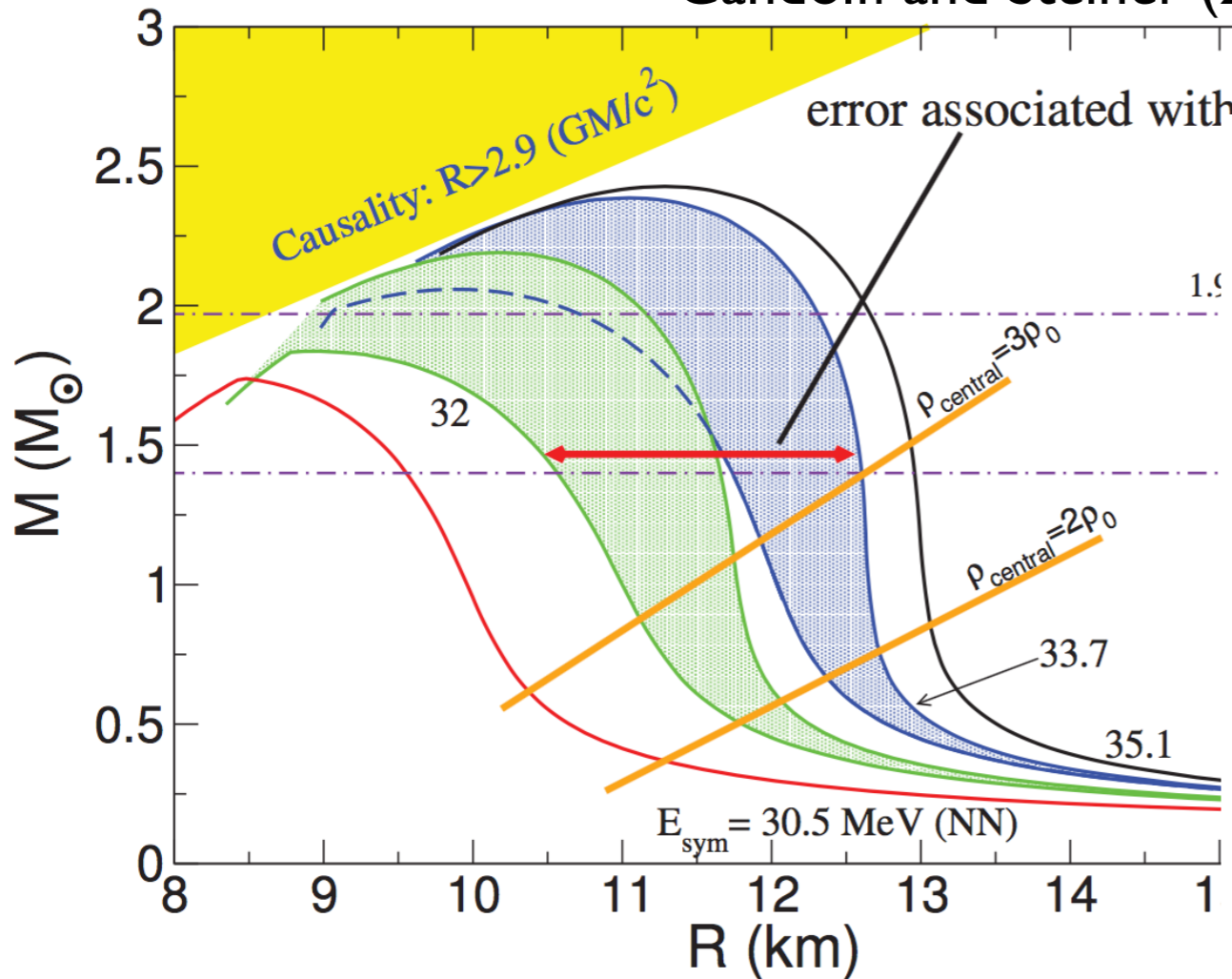
$$E_{sym}(\rho) = E_{sym}(\rho_0) + \frac{L}{3} \left(\frac{\rho - \rho_0}{\rho_0} \right) + \frac{K_{sym}}{18} \left(\frac{\rho - \rho_0}{\rho_0} \right)^2$$

$$L = 3\rho_0 \frac{\partial E_{sym}(\rho)}{\partial \rho} \Big|_{\rho=\rho_0} \quad K_{sym} = 9\rho_0^2 \frac{\partial^2 E_{sym}(\rho)}{\partial \rho^2} \Big|_{\rho=\rho_0}.$$

Nuclear Symmetry Energy

J.M. Lattimer / Nuclear Physics A 928 (2014) 276–295

Gandolfi and Steiner (2016)



$$\frac{E}{A} = E(\rho, \delta = 0) + E_{sym}(\rho)\delta^2 + \mathcal{O}(\delta^4) + \dots,$$

$$E_{sym}(\rho) = E_{sym}(\rho_0) + \frac{L}{3} \left(\frac{\rho - \rho_0}{\rho_0} \right) + \frac{K_{sym}}{18} \left(\frac{\rho - \rho_0}{\rho_0} \right)^2$$

isospin-asymmetry energy slope parameter L

$$L = 3\rho_0 \left. \frac{\partial E_{sym}(\rho)}{\partial \rho} \right|_{\rho=\rho_0} \quad K_{sym} = 9\rho_0^2 \left. \frac{\partial^2 E_{sym}(\rho)}{\partial \rho^2} \right|_{\rho=\rho_0}.$$

[D] Y.Lim and J. Holt, PRL.121.062701 (arXiv:1803.02803v2)

Bayesian Analysis

Bayes' theorem

The likelihood could be the function of errors

Posterior $p(\theta|d) = \frac{p(d|\theta)}{p(d)} \cdot p(\theta)$

Prior choices can influence results

$$= \frac{p(d|\theta)}{\int p(d|\theta)p(\theta)d\theta} \cdot p(\theta)$$

The evidence is unimportant for parameter estimation (but not model selection !)

$$p(\theta|d) \sim p(d|\theta)p(\theta)$$

Prior, $p(\theta)$: the distribution of the parameter(s) before any data is observed

Likelihood, $p(d|\theta)$: the distribution of the observed data conditional on its parameters

Posterior, $p(\theta|d)$: the distribution of the parameter(s) after taking into account the observed data

Model evidence, $p(d)$:the distribution of the observed data **marginalized** over the parameter(s)

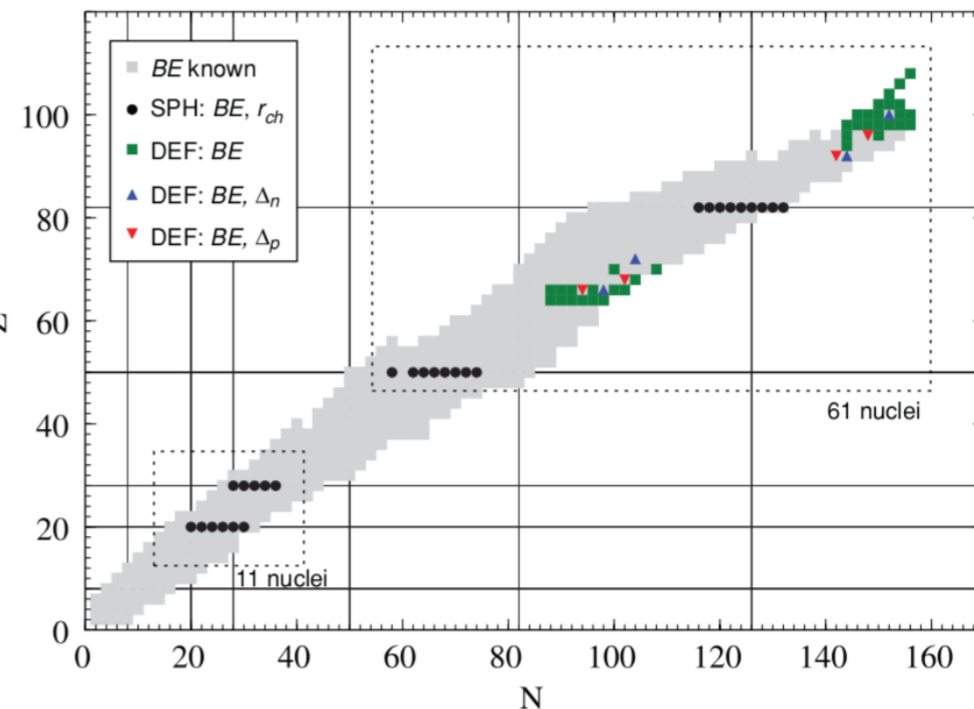
Bayesian Analysis for KIDS - 1st approach

$$L \sim \exp(-\Sigma(d_i - m_i)^2 / 2\sigma^2)$$

$$p(\theta|d) = \frac{p(\theta|d)}{p(d)} \cdot p(\theta)$$

$$\mathcal{E}(\rho, \delta) = \frac{E(\rho, \delta)}{A} = \mathcal{T}(\rho, \delta) + \Sigma_{i=0}^{N-1} c_i(\delta) \rho^{1+i/3}$$

PHYSICAL REVIEW C **82**, 024313 (2010)



+ $\langle R_{ch}^2 \rangle$

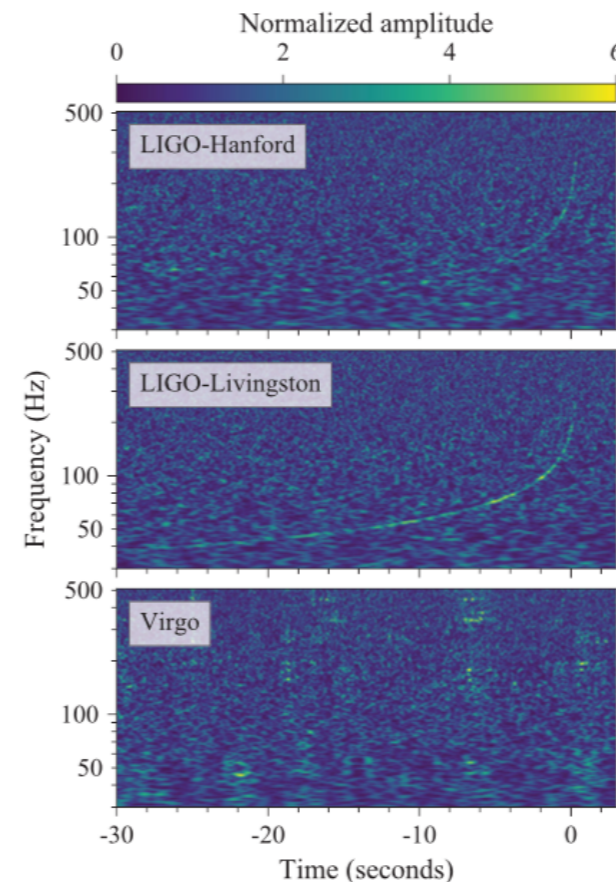
Do MCMC (Metropolis-Hasting algorithm)

Bayesian Analysis for KIDS - 2nd approach

$$L \sim \exp(-\Sigma(d_i - m_i)^2 / 2\sigma^2)$$

$$p(\theta|d) = \frac{p(\theta|d)}{p(d)} \cdot p(\theta)$$

$$\mathcal{E}(\rho, \delta) = \frac{E(\rho, \delta)}{A} = \mathcal{T}(\rho, \delta) + \sum_{i=0}^{N-1} c_i(\delta) \rho^{1+i/3}$$



+ EM Obs.
(LMXB, etc.)

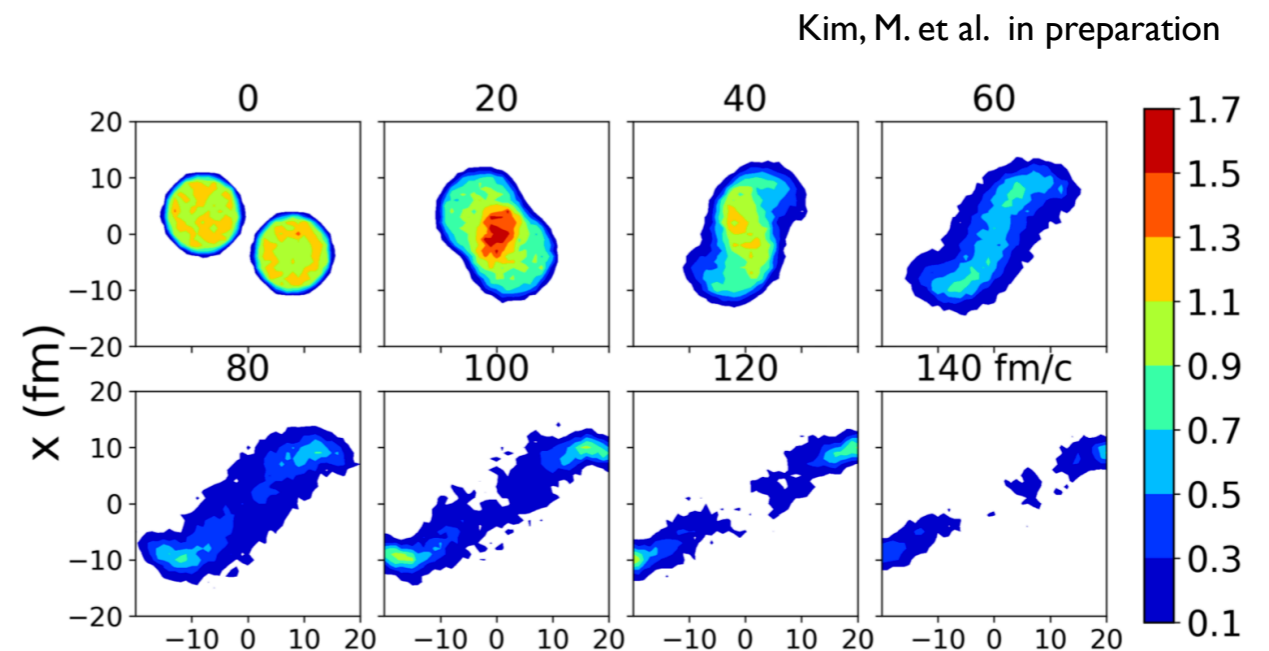
Do MCMC (Metropolis-Hasting algorithm)

Bayesian Analysis for KIDS - 3rd approach

$$L \sim \exp(-\Sigma(d_i - m_i)^2 / 2\sigma^2)$$

$$p(\theta|d) = \frac{p(\theta|d)}{p(d)} \cdot p(\theta)$$

$$\mathcal{E}(\rho, \delta) = \frac{E(\rho, \delta)}{A} = \mathcal{T}(\rho, \delta) + \Sigma_{i=0}^{N-1} c_i(\delta) \rho^{1+i/3}$$

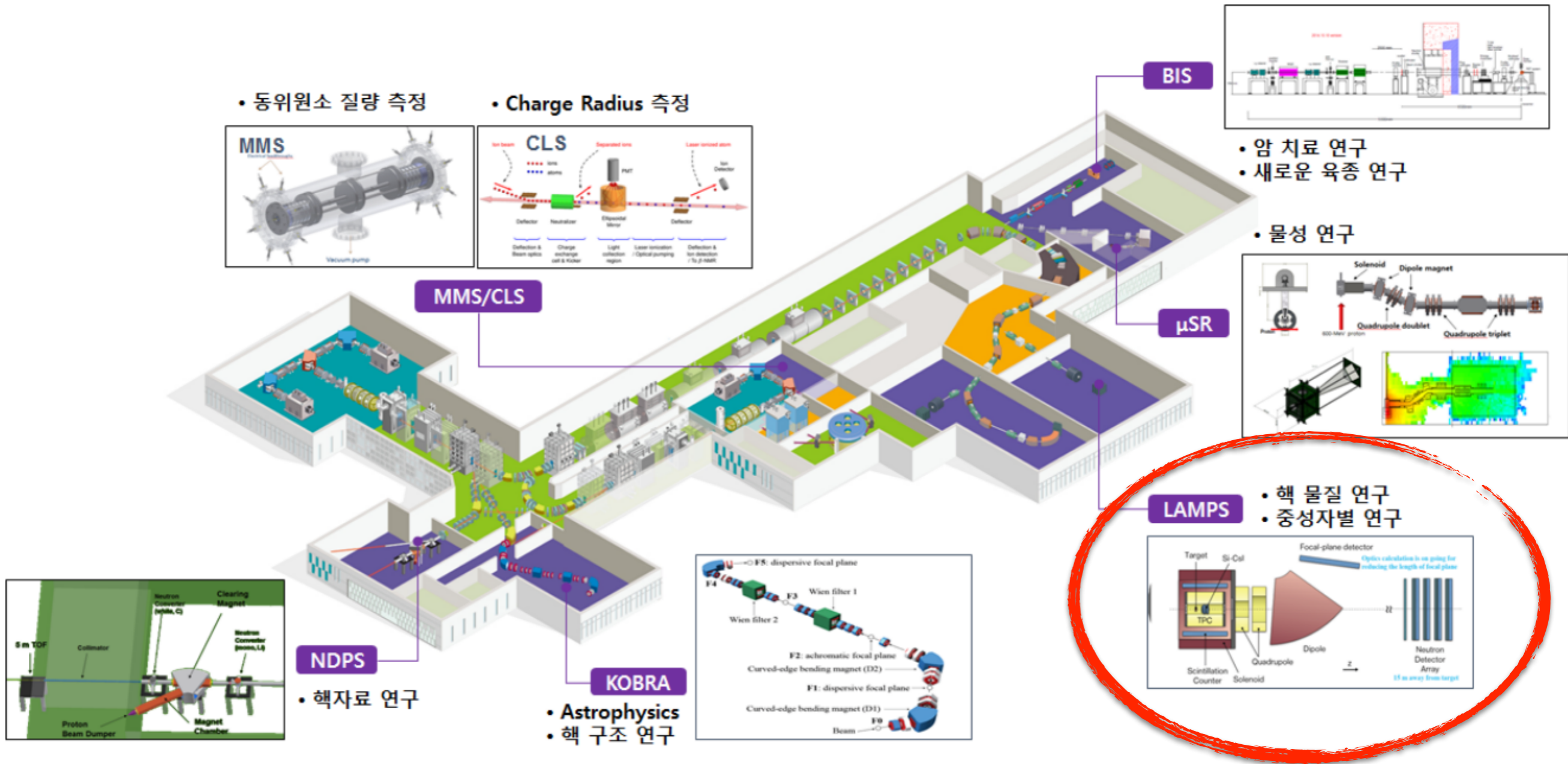


¹⁹⁷Au+¹⁹⁷Au@ 100 AMeV, b=7fm

Do MCMC (Metropolis-Hasting algorithm)

RAON - Rare Isotope Accelerator

Experimental Systems



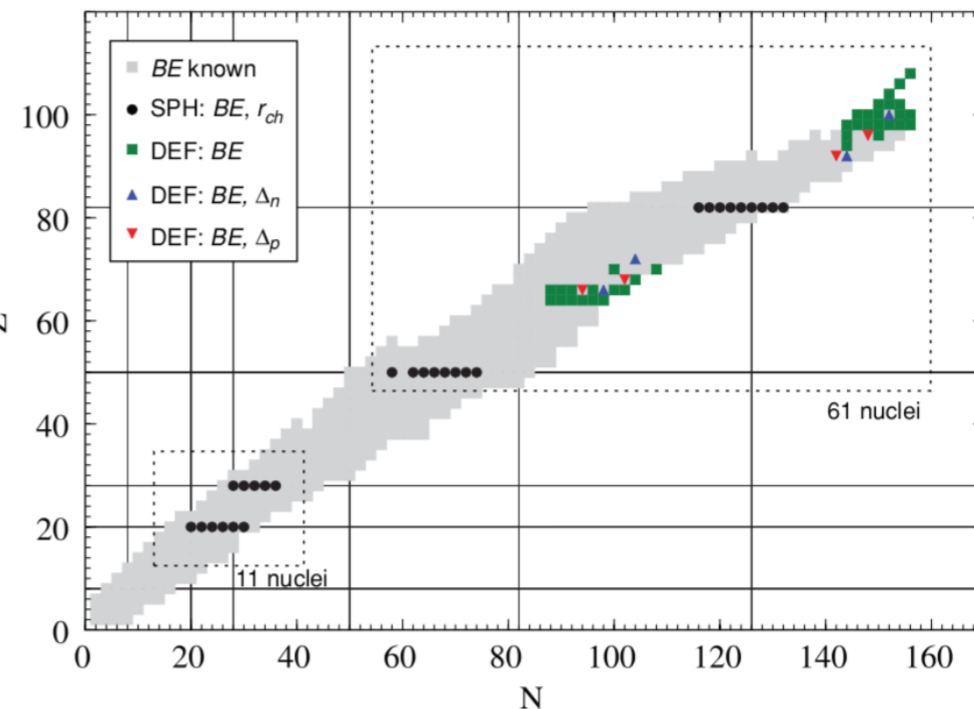
Bayesian Analysis for KIDS - 1st approach

$$L \sim \exp(-\Sigma(d_i - m_i)^2 / 2\sigma^2)$$

$$p(\theta|d) = \frac{p(\theta|d)}{p(d)} \cdot p(\theta)$$

$$\mathcal{E}(\rho, \delta) = \frac{E(\rho, \delta)}{A} = \mathcal{T}(\rho, \delta) + \Sigma_{i=0}^{N-1} c_i(\delta) \rho^{1+i/3}$$

PHYSICAL REVIEW C **82**, 024313 (2010)



Do MCMC (Metropolis-Hasting algorithm)

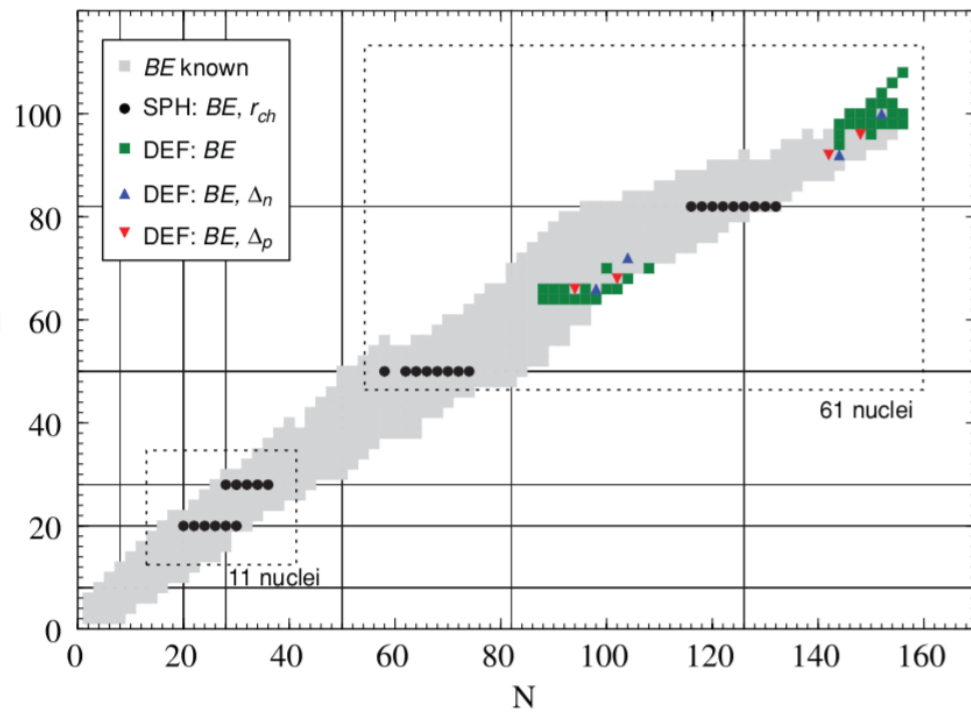
Bayesian Analysis for KIDS - details (I)

$$L \sim \exp(-\Sigma(d_i - m_i)^2 / 2\sigma^2)$$

$$p(\theta|d) = \frac{p(\theta|d)}{p(d)} \cdot p(\theta)$$

Previous studies

PHYSICAL REVIEW C **82**, 024313 (2010)



+ $\langle R_{ch}^2 \rangle$

H. Gil, et al. PRC, 100, 014312 (2019)

Parameter	S3b	S4a	S4b	S4c
t_0 (MeV fm ³)	-1772.04	-1807.17	-1723.84	-1640.50
y_0 (MeV fm ³)	-127.52	-162.65	-79.32	4.02
t_1 (MeV fm ⁵)	275.72	262.17	288.94	303.28
t_2 (MeV fm ⁵)	-161.50	-167.94	-154.90	-146.98
t_{31} (10 ⁴ MeV fm ⁴)	1.222	1.338	1.062	0.7855
y_{31} (10 ⁴ MeV fm ⁴)	-1.197	-1.081	-1.357	-1.633
t_{32} (MeV fm ⁵)	571.0	-1310.7	3252.4	8043.0
y_{32} (10 ⁴ MeV fm ⁵)	2.949	2.704	3.274	3.818
t_{33} (MeV fm ⁶)	-	1317.2	-1807.8	-4932.8
y_{33} (10 ⁴ MeV fm ⁶)	-2.296	-2.164	-2.476	-2.789
ζ	0.1106	0.1281	0.0931	0.0729
W_0 (MeV fm ⁵)	108.35	106.79	109.88	111.55

Bayesian Analysis for KIDS - details (2)

H. Gil, et al. PRC, 100, 014312 (2019)

Nucleus	Binding energy per nucleon (E/A) (MeV)					Charge radius (R_c) (fm)				
	Expt.	S3b	S4a	S4b	S4c	Expt.	S3b	S4a	S4b	S4c
^{40}Ca	8.5513*	8.5565 (0.060%)	8.5579 (0.078%)	8.5544 (0.037%)	8.5512 (0.001%)	3.4776*	3.4781 (0.014%)	3.4799 (0.066%)	3.4758 (0.052%)	3.4720 (0.161%)
^{48}Ca	8.6667*	8.6564 (0.120%)	8.6569 (0.113%)	8.6558 (0.126%)	8.6549 (0.136%)	3.4771*	3.4867 (0.277%)	3.4882 (0.319%)	3.4847 (0.220%)	3.4813 (0.122%)
^{208}Pb	7.8675*	7.8809 (0.172%)	7.8816 (0.179%)	7.8800 (0.160%)	7.8783 (0.138%)	5.5012*	5.4887 (0.228%)	5.4901 (0.201%)	5.4870 (0.259%)	5.4840 (0.313%)
^{16}O	7.9762	7.8684 (1.35%)	7.8675 (1.36%)	7.8686 (1.35%)	7.8678 (1.36%)	2.6991	2.7618 (2.322%)	2.7643 (2.41%)	2.7587 (2.209%)	2.7541 (2.036%)
^{28}O	–	6.0646	6.0640	6.0650	6.0649	–	2.8371	2.8384	2.8351	2.8315
^{60}Ca	–	7.6561	7.6567	7.6552	7.6535	–	3.6465	3.6478	3.6445	3.6411
^{90}Zr	8.7100	8.7328 (0.263%)	8.7345 (0.281%)	8.7309 (0.241%)	8.7282 (0.209%)	4.2694	4.2476 (0.510%)	4.2488 (0.482%)	4.2459 (0.550%)	4.2428 (0.622%)
^{132}Sn	8.3549	8.3563 (0.017%)	8.3559 (0.013%)	8.3565 (0.020%)	8.3565 (0.020%)	4.7093	4.7089 (0.009%)	4.7100 (0.015%)	4.7072 (0.044%)	4.7044 (0.103%)

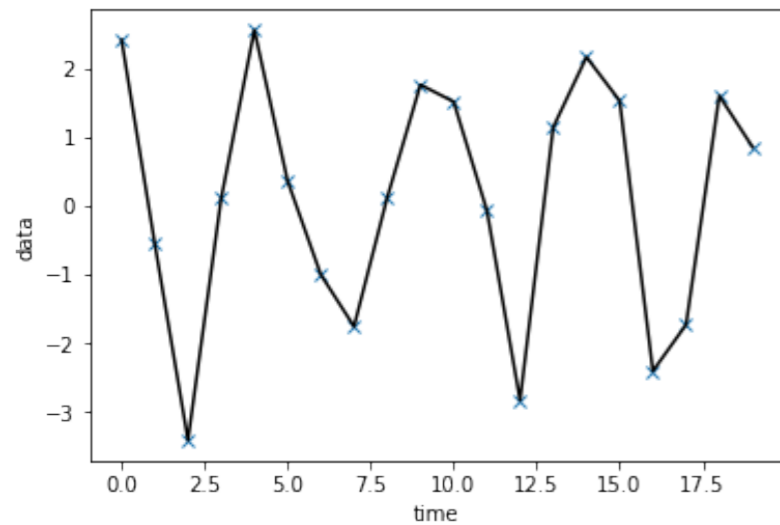
Skyrem-Hatree-Fock Calculation

My Hope is

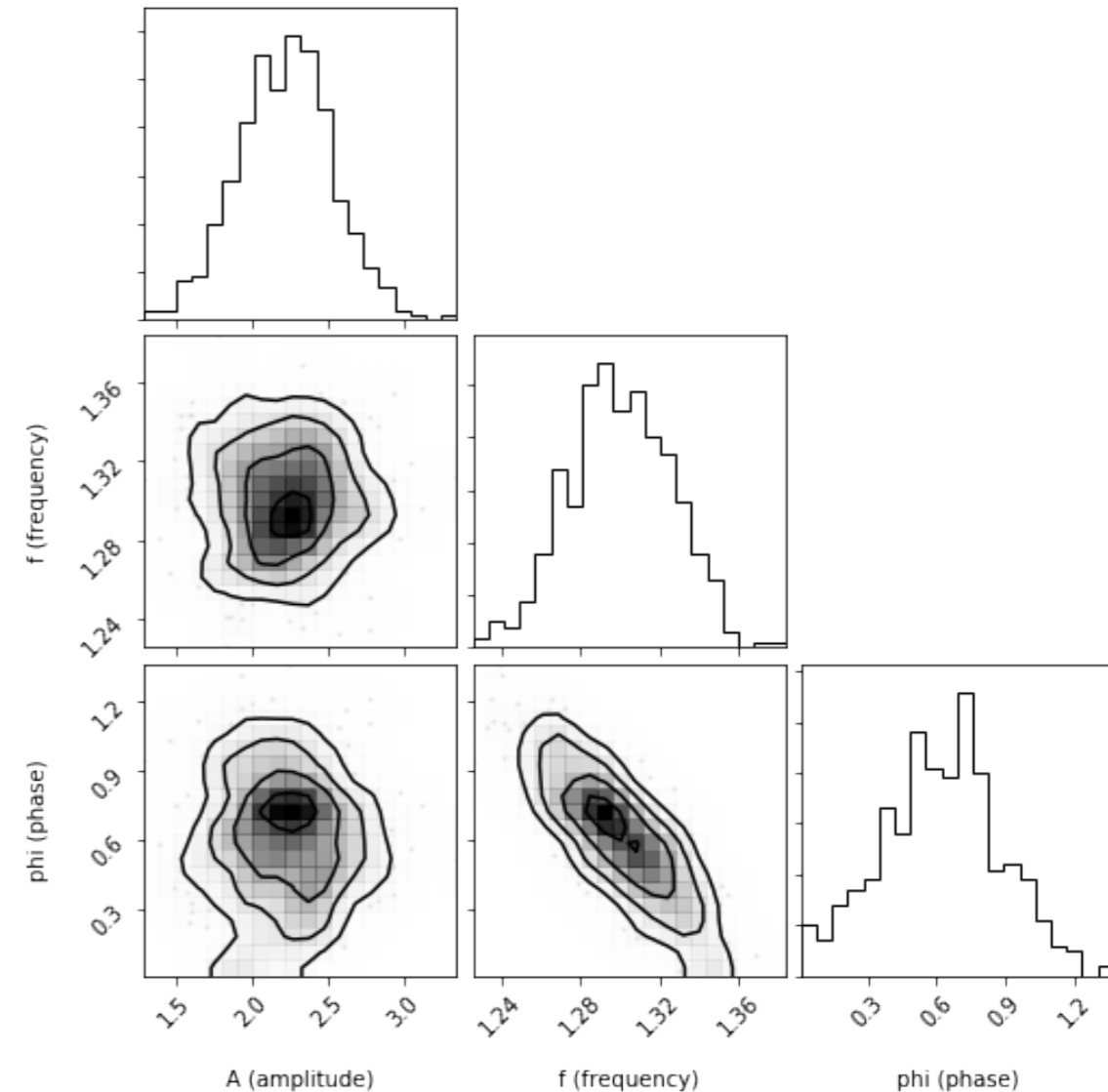
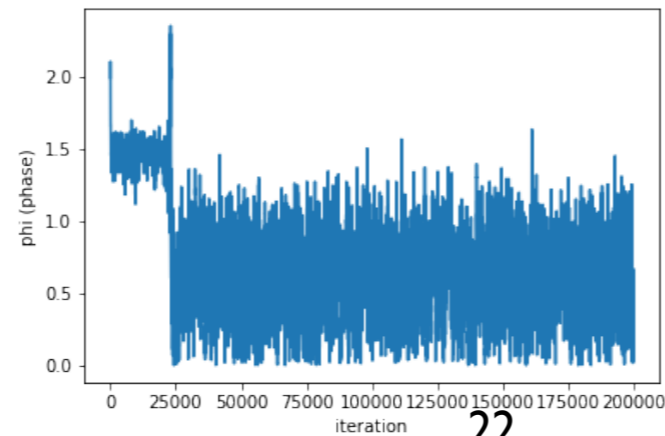
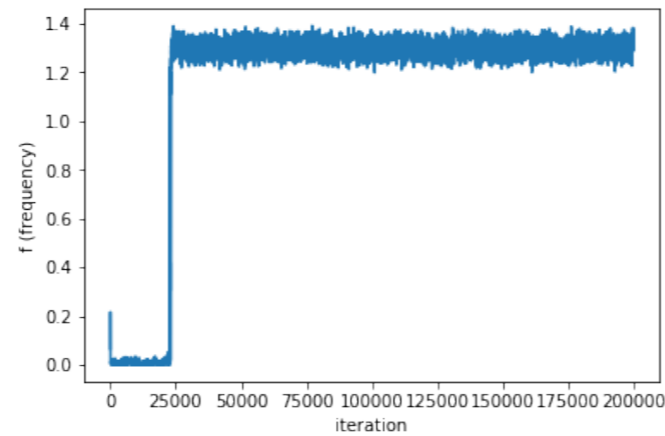
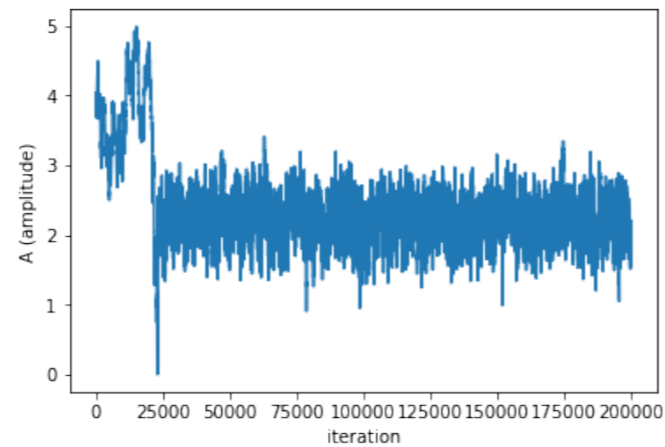
1. A MCMC example : https://github.com/gw-odw/odw-2018/blob/master/parameter_estimation/IntroToMCMC.ipynb

$$d = A \cdot \cos(f \cdot t + \phi)$$

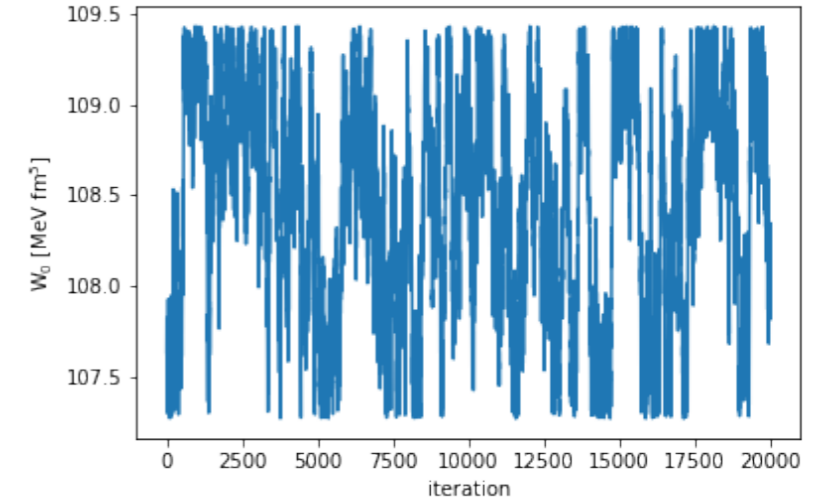
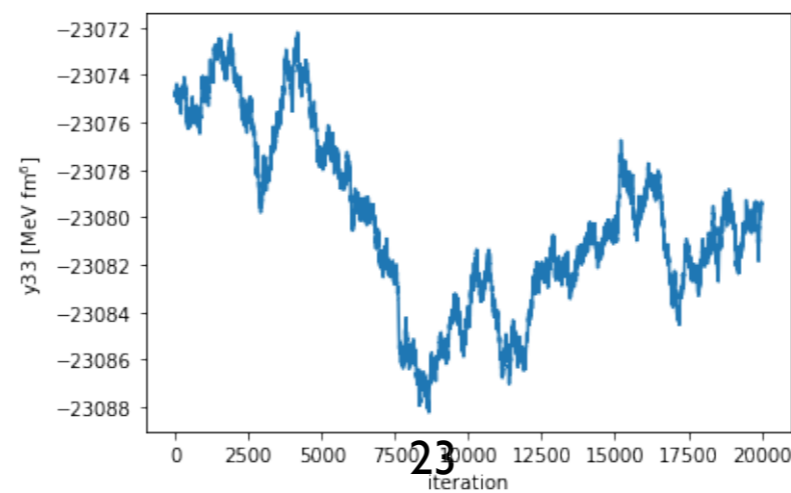
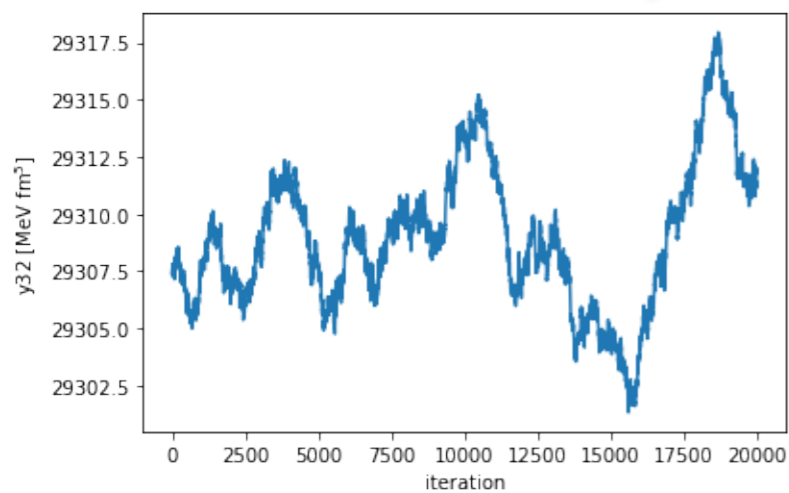
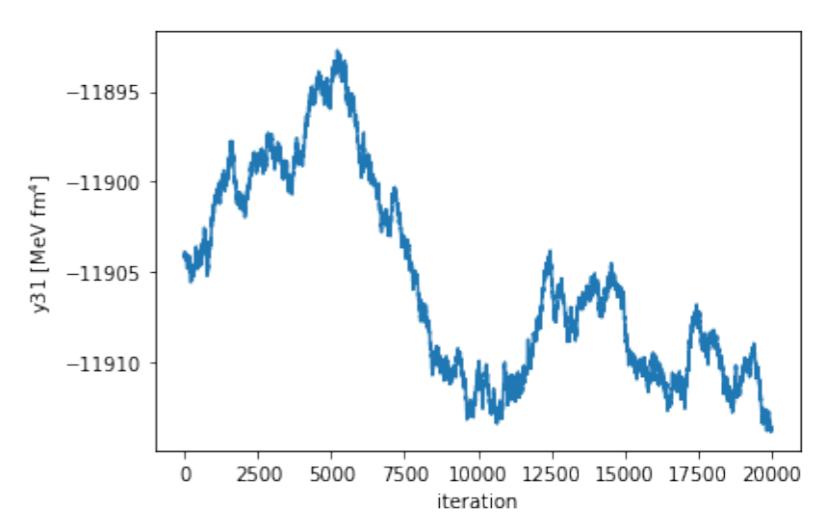
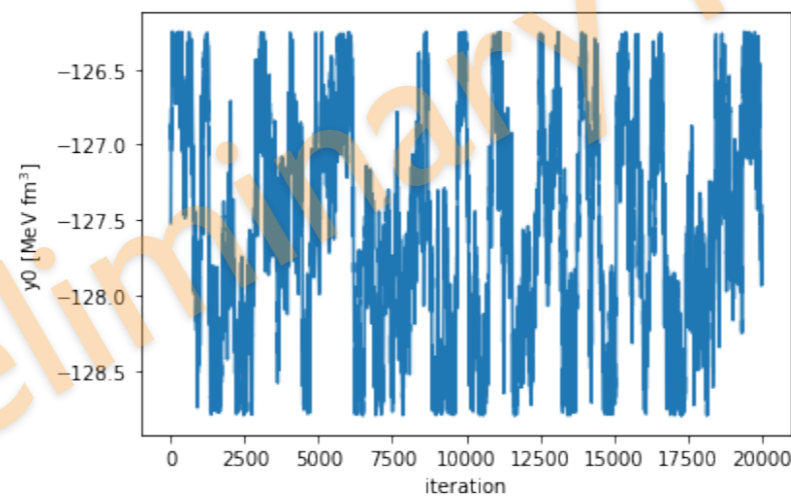
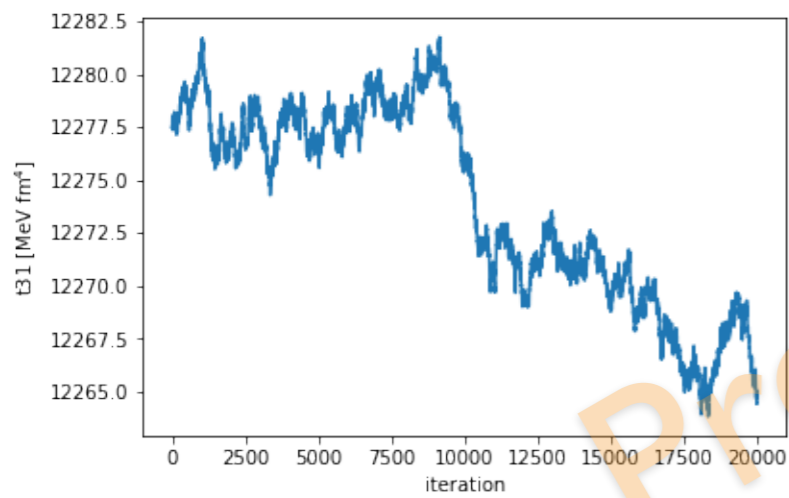
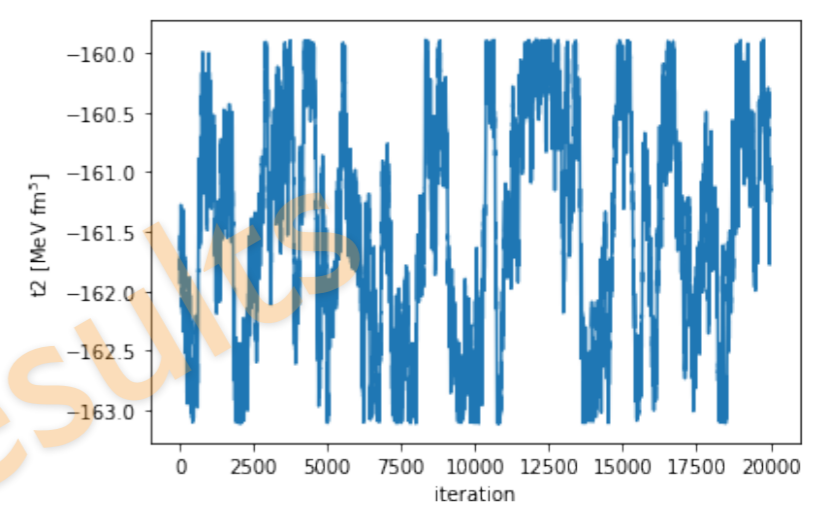
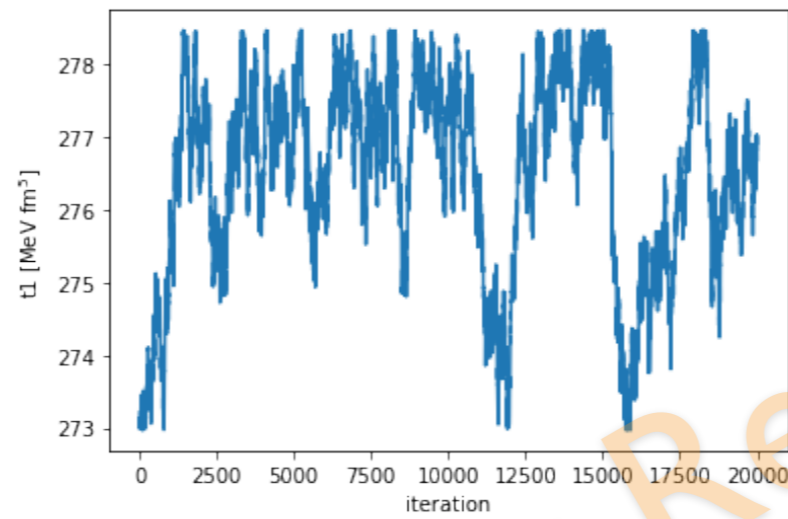
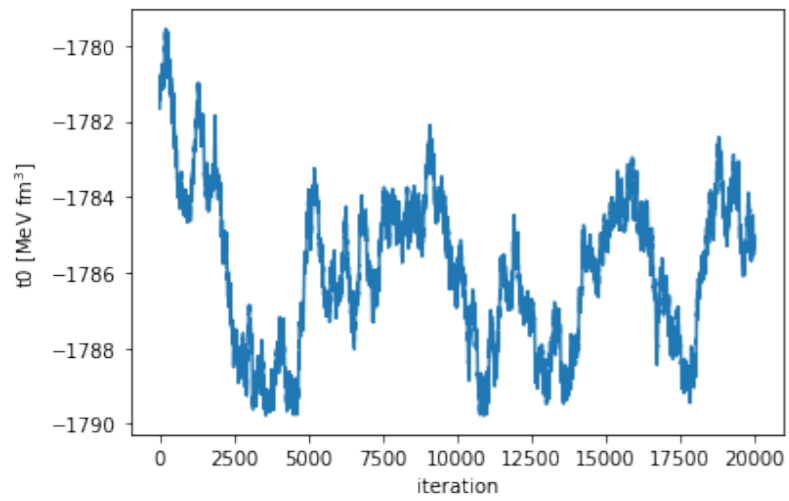
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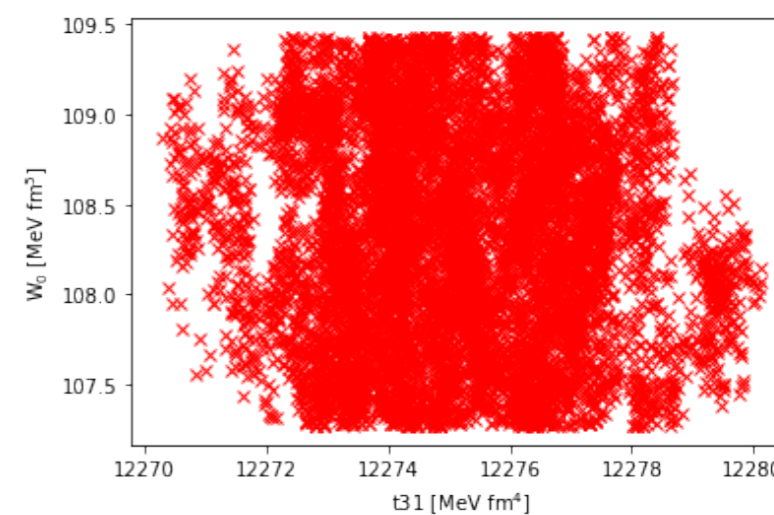
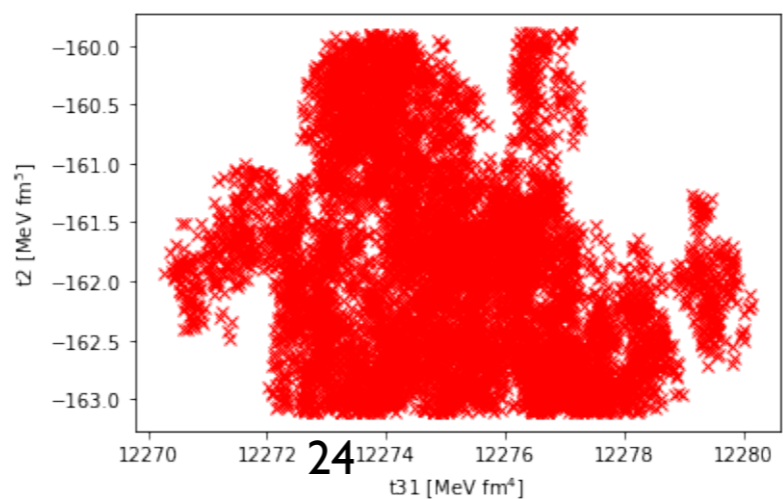
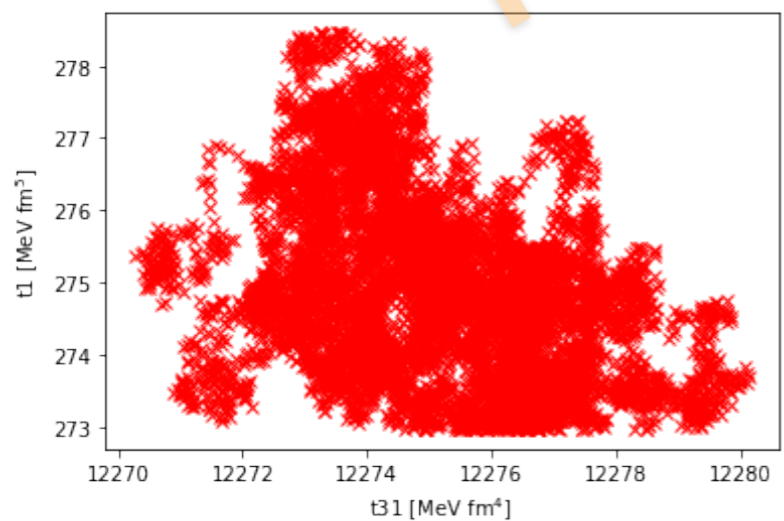
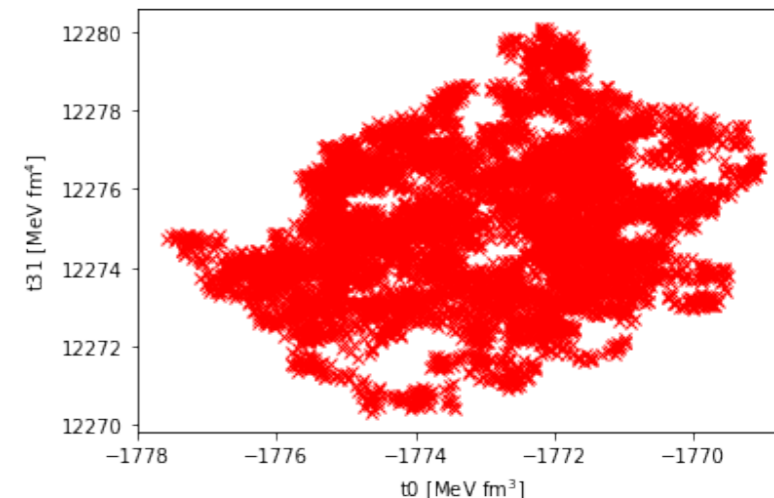
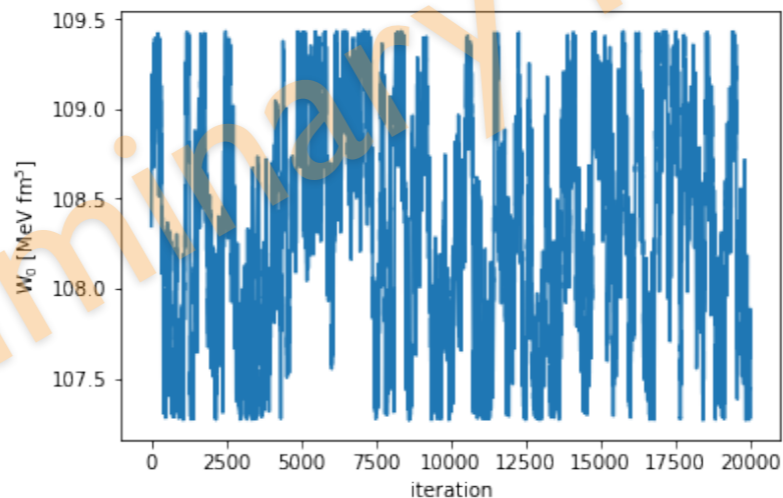
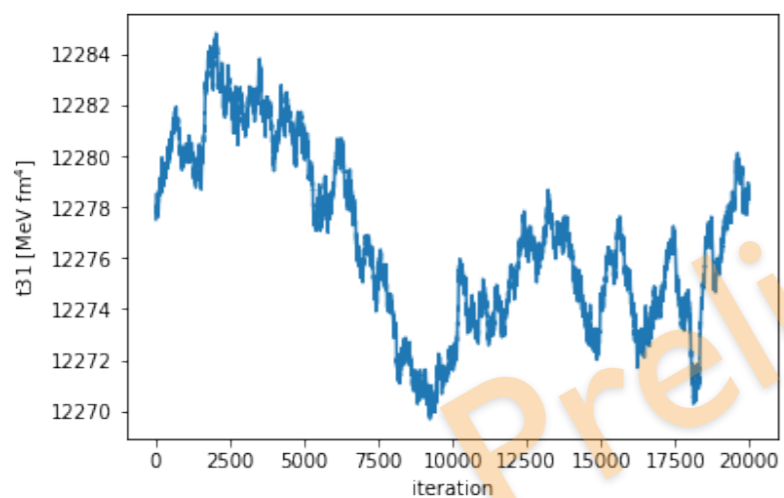
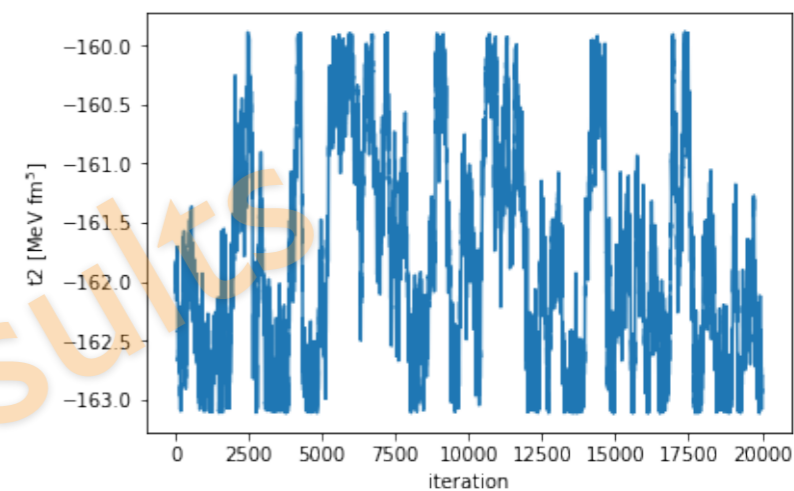
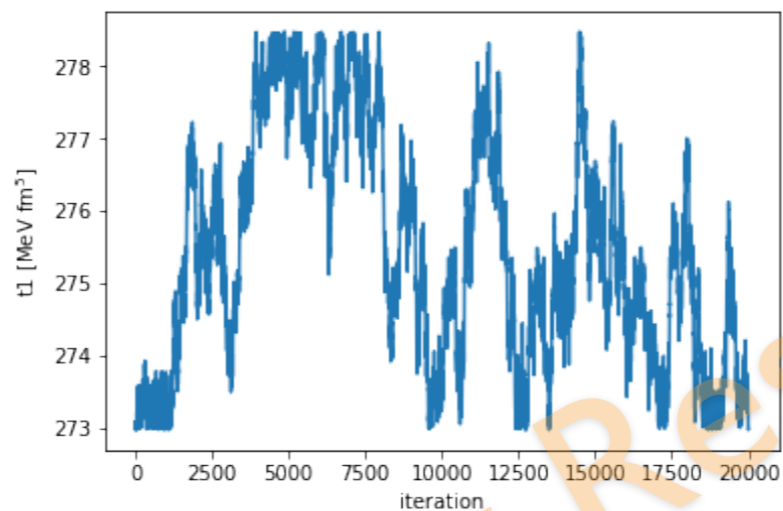
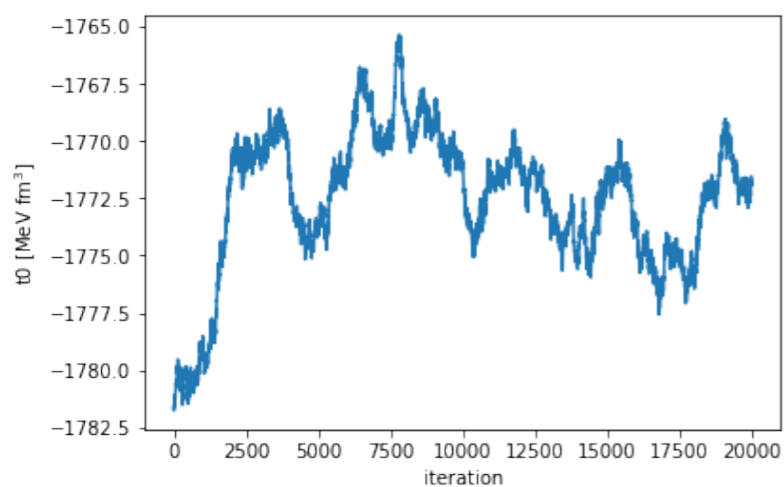
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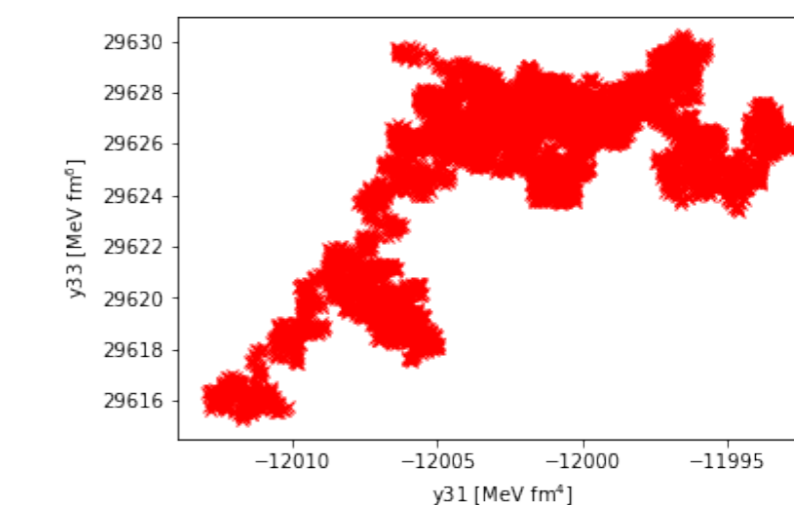
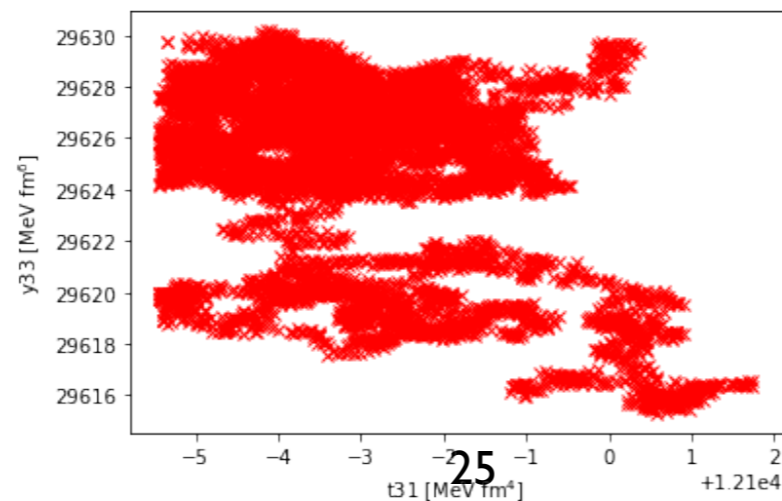
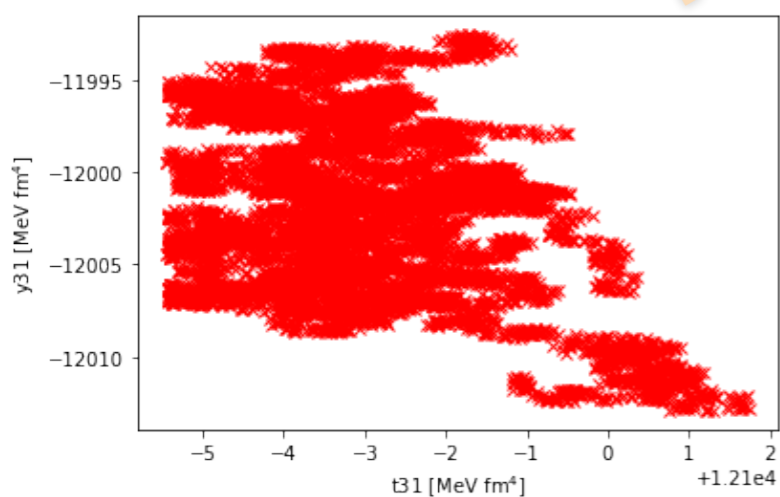
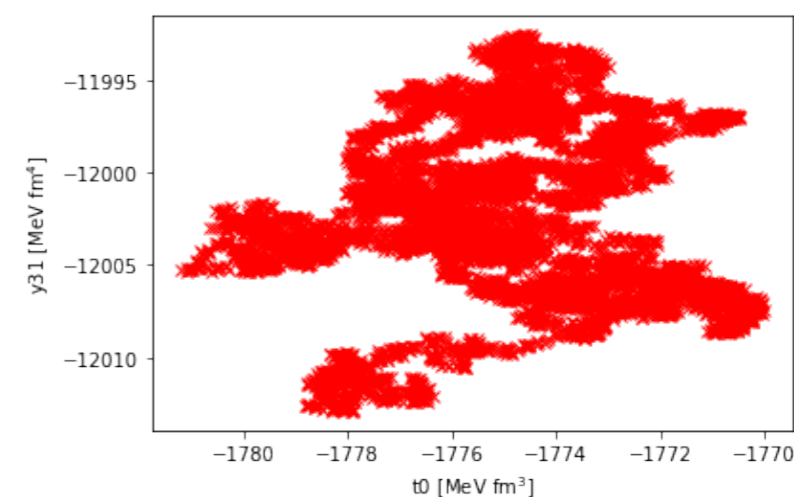
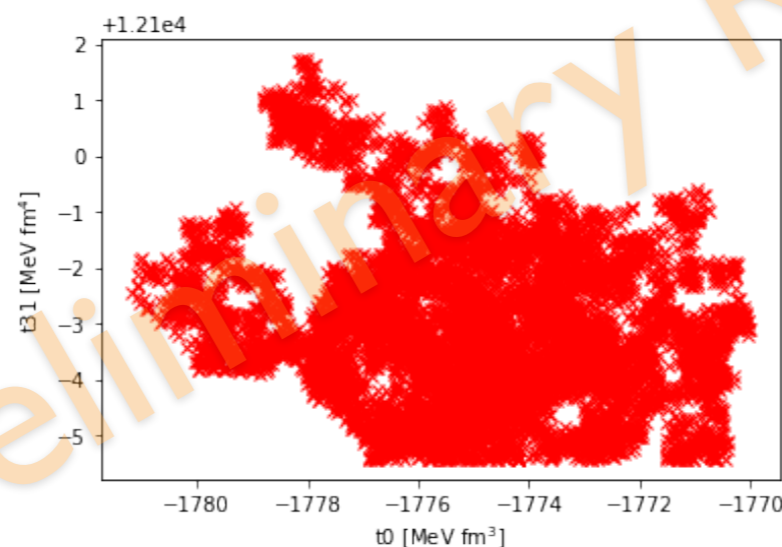
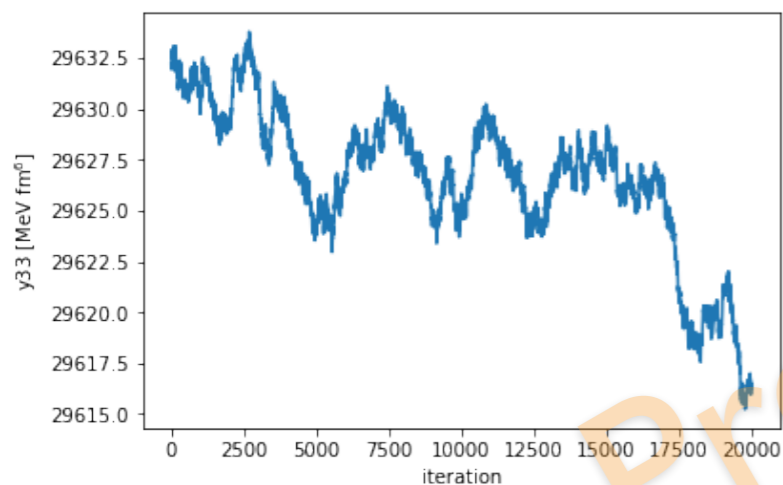
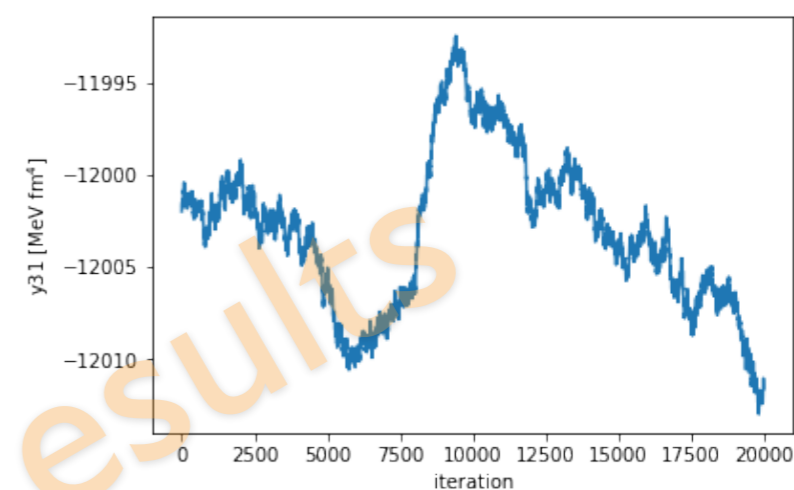
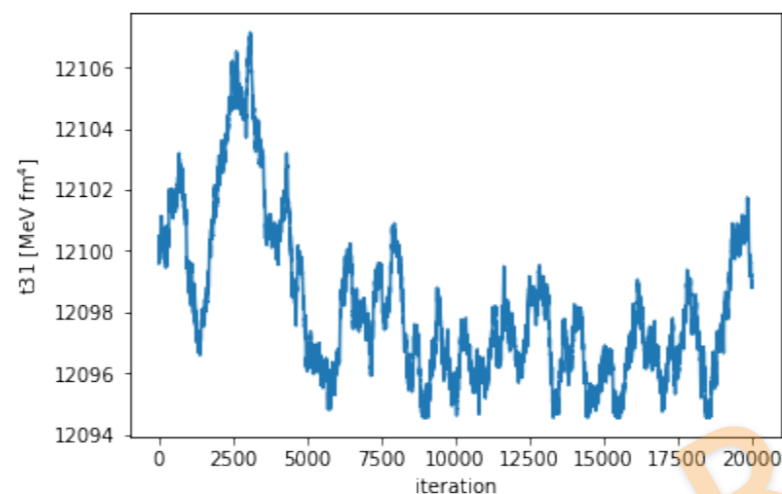
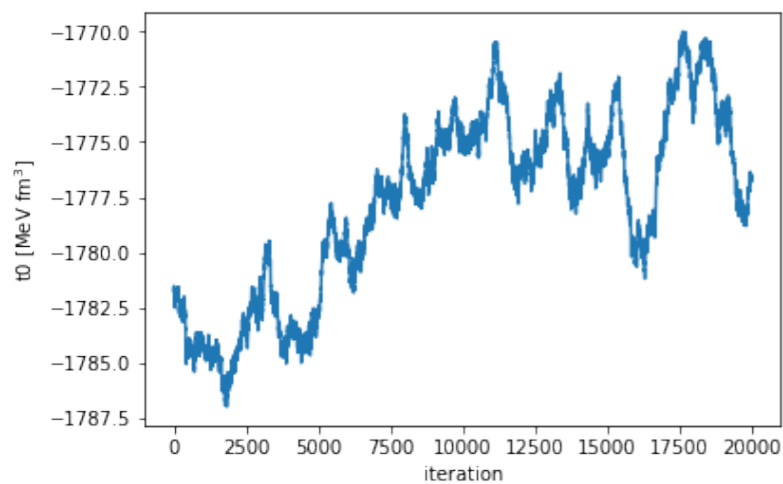
But, Reality is



Variables : t_0 , t_1 , t_2 , t_3 , W_0



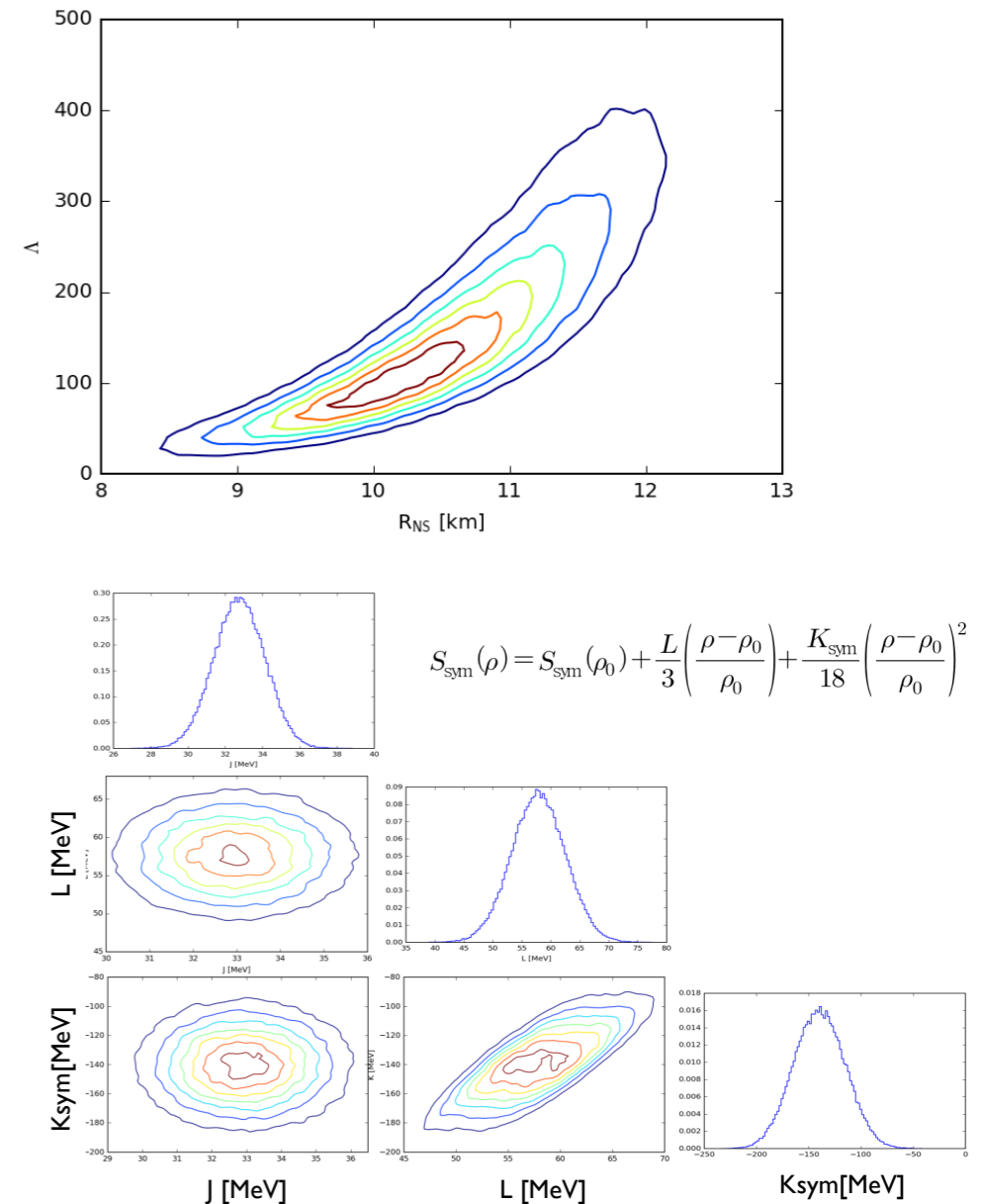
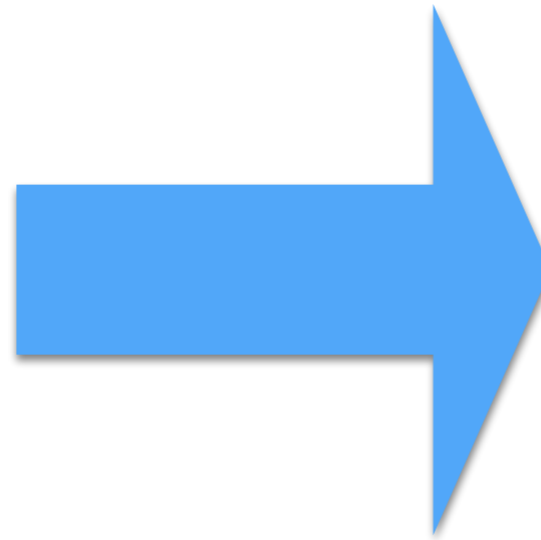
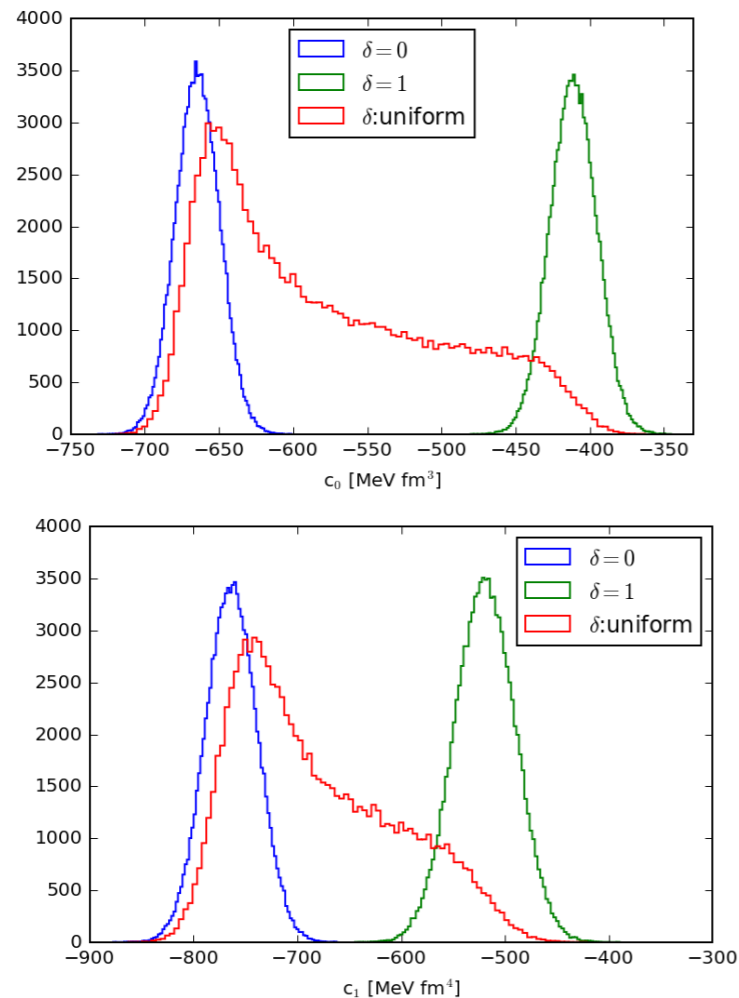
Variables : t_0 , t_{31} , y_{31} , y_{33}



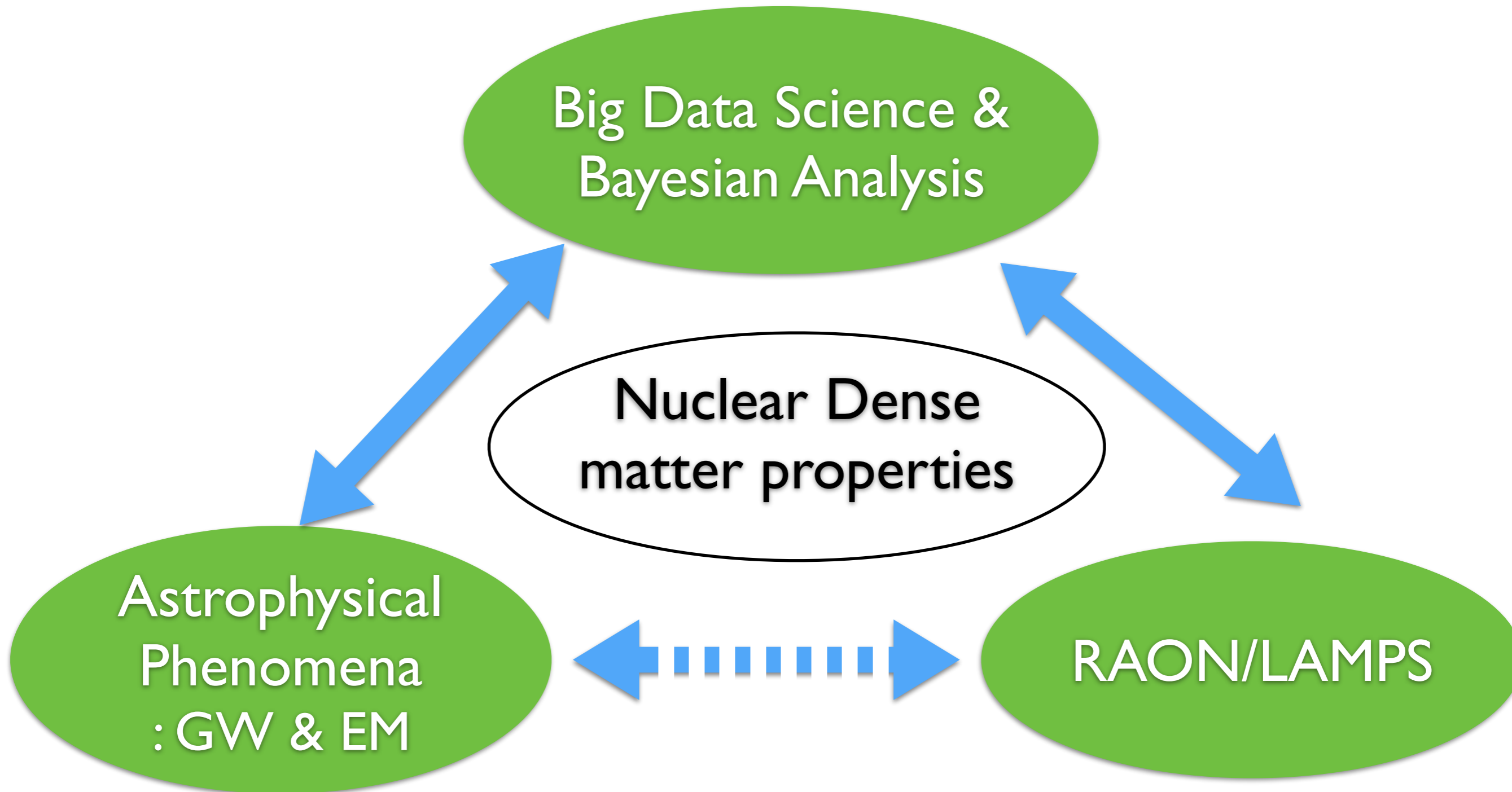
Future Plan

I. Bayesian Analysis on GWDA as well as RAON/LAMPS exp. + DJBUU

- To Do better !!
- MCMC code (C++ based) development for various applications



What I want to do with Bayesian ...



Thank you for your attention.