

<sup>1</sup>Hebrew University of Jerusalem



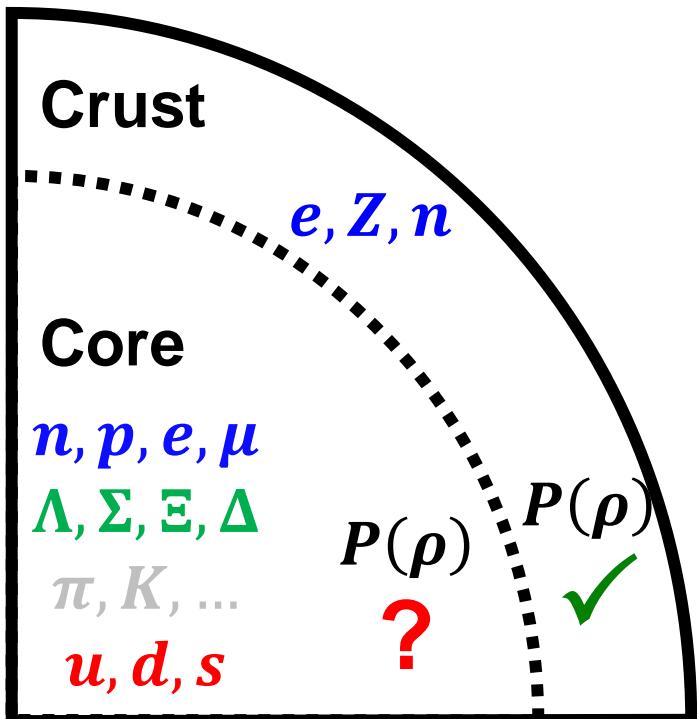
<sup>2</sup>Ioffe Institute

# Self-similarities of Equations of State and $M - R$ curves of Neutron Stars

**Dima Ofengeim<sup>1,2</sup>, P. Shternin<sup>2</sup>, T. Piran<sup>1</sup>**

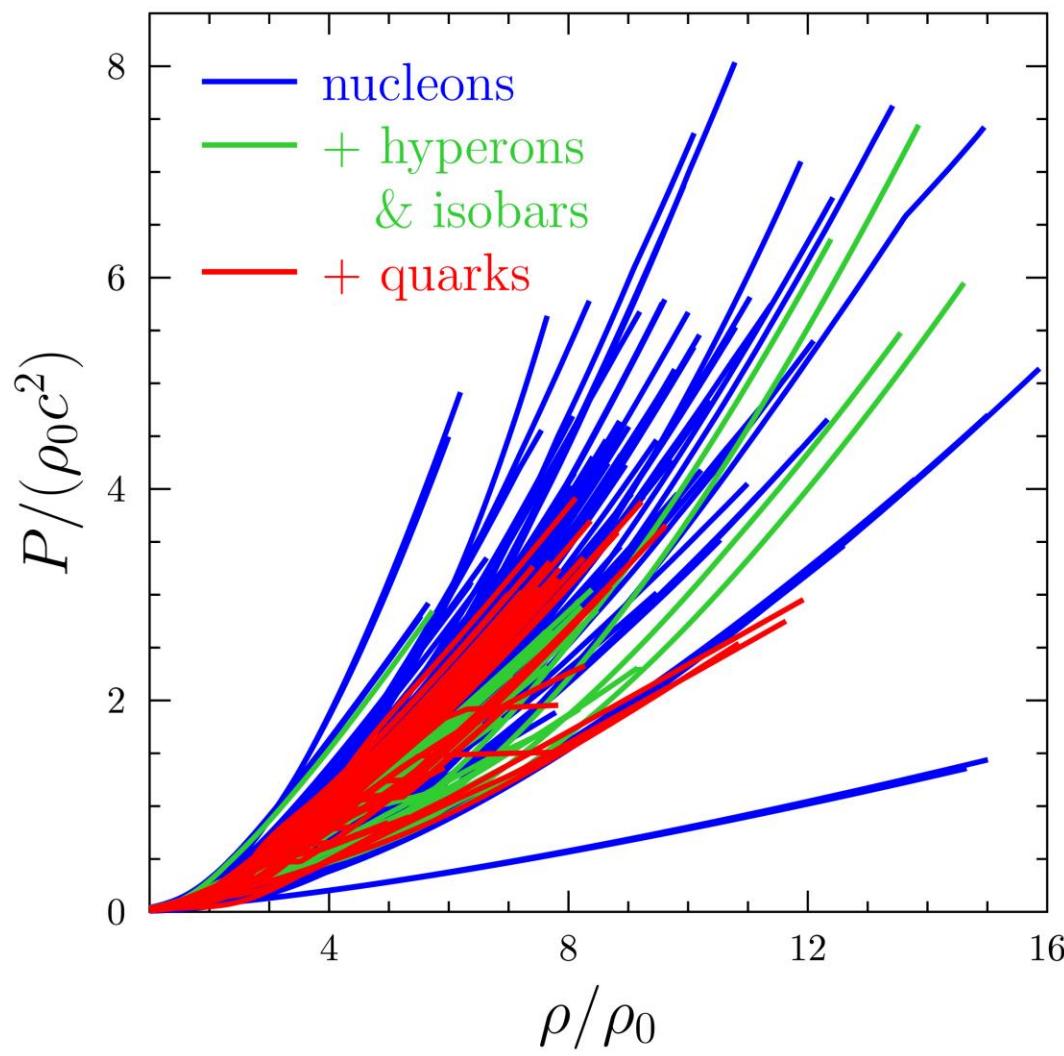
# NS Equations of State

- Equation of state = EoS
- Cold degenerate matter  
 $T < 10^{10}$  K,  $T_F \sim 10^{12}$  K



$$\rho_0 = 2.8 \times 10^{14} \text{ g/cm}^3$$
$$\sim 0.5\rho_0$$

- $P(\rho), n(\rho)$ , composition( $\rho$ ), ...



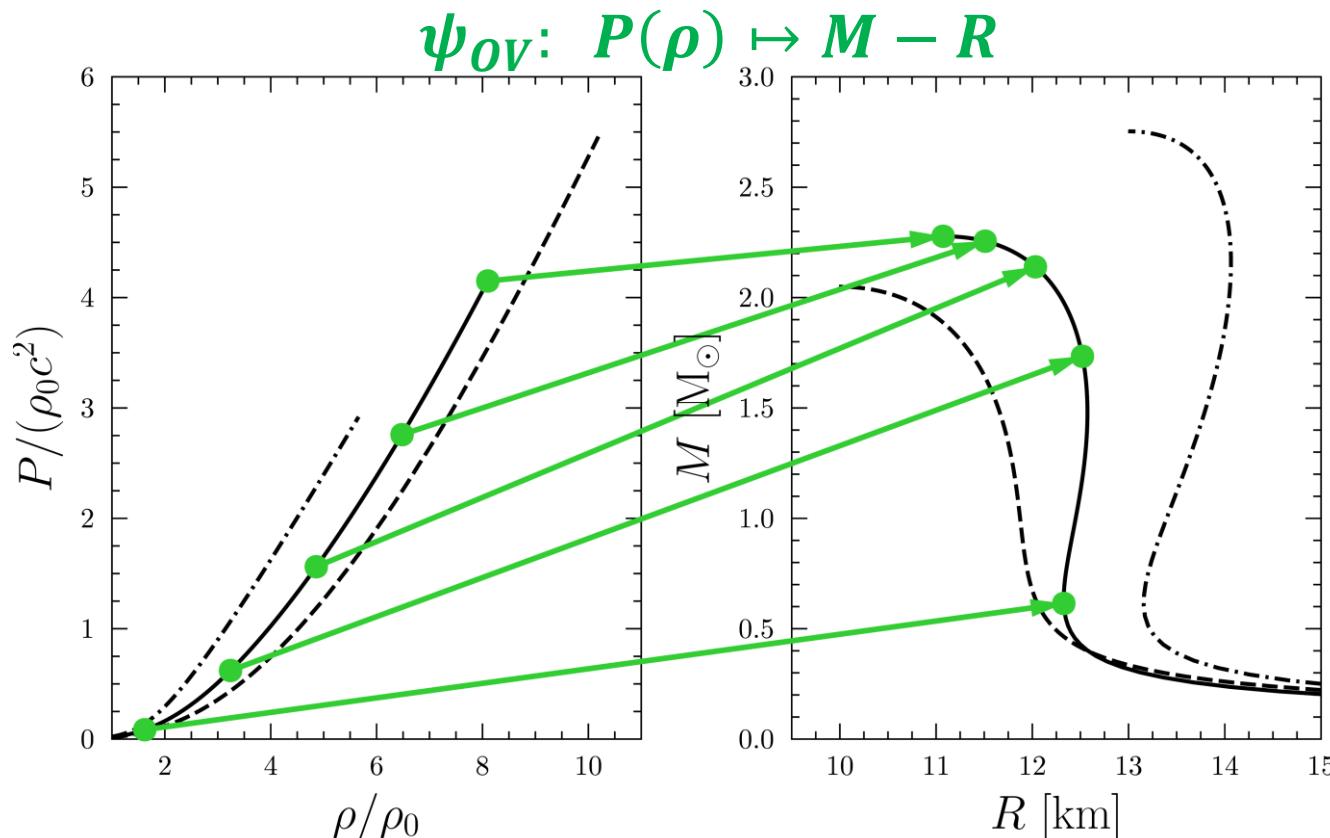
# Oppenheimer-Volkoff Mapping

$$\frac{dP}{dr} = -\frac{Gm\rho}{r^2} \frac{\left(1 + \frac{P}{\rho c^2}\right)\left(1 + \frac{4\pi r^3 P}{mc^2}\right)}{1 - \frac{2Gm}{rc^2}}$$

➤ **GR hydrostatics**  
➤ ~~Rotation~~

Tolman (1939); Oppenheimer & Volkoff (1939)

$$\frac{dm}{dr} = 4\pi r^2 \rho$$



# Oppenheimer-Volkoff Mapping

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Tolman (1939); Oppenheimer & Volkoff (1939)

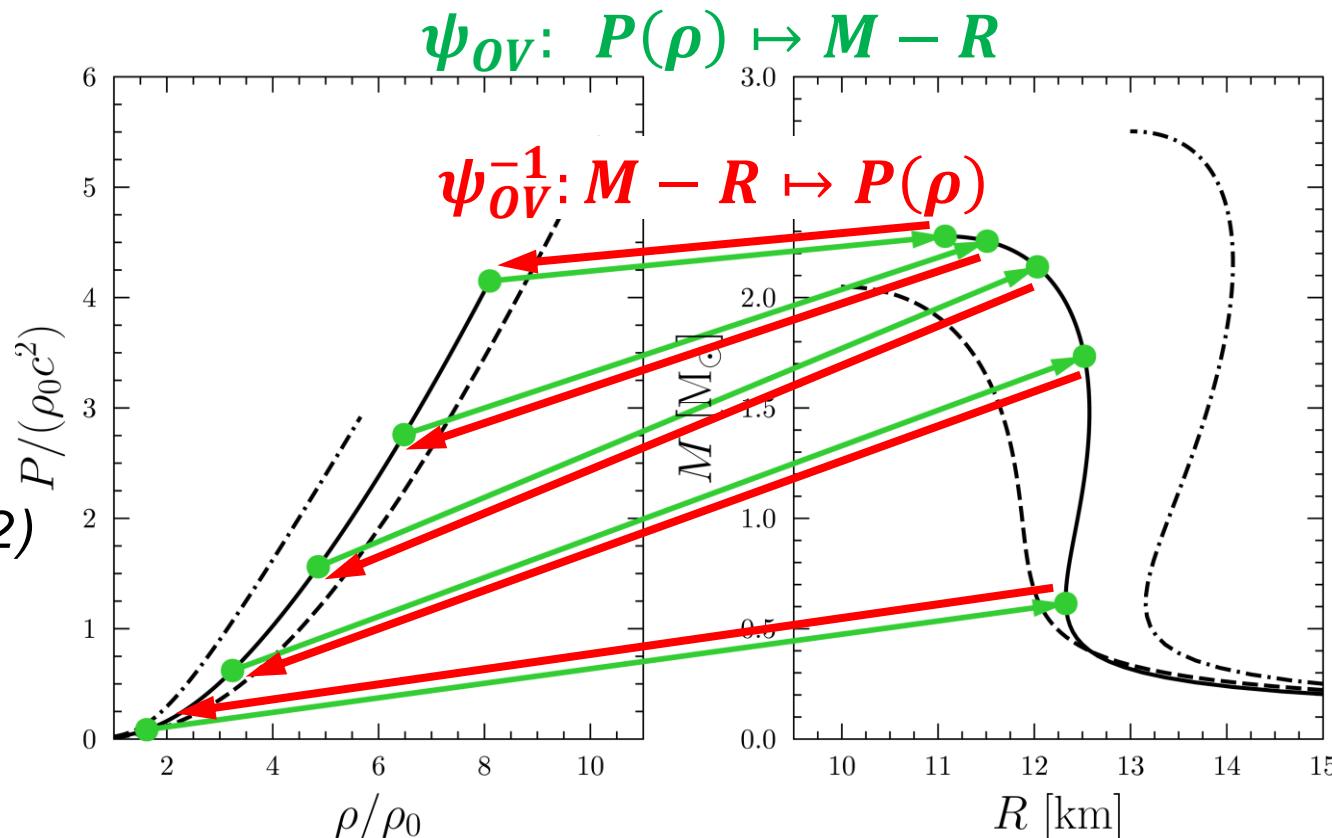
$$\frac{dm}{dr} = 4\pi r^2 \rho$$

- Lindblom (1992):

$\exists \psi_{OV}^{-1}$

➤ numerics

➤ neural networks  
(Soma+ JCAP 2022)



# Oppenheimer-Volkoff Mapping

$$\frac{dP}{dr} = -\frac{Gm\rho}{r^2} \frac{\left(1 + \frac{P}{\rho c^2}\right)\left(1 + \frac{4\pi r^3 P}{mc^2}\right)}{1 - \frac{2Gm}{rc^2}}$$

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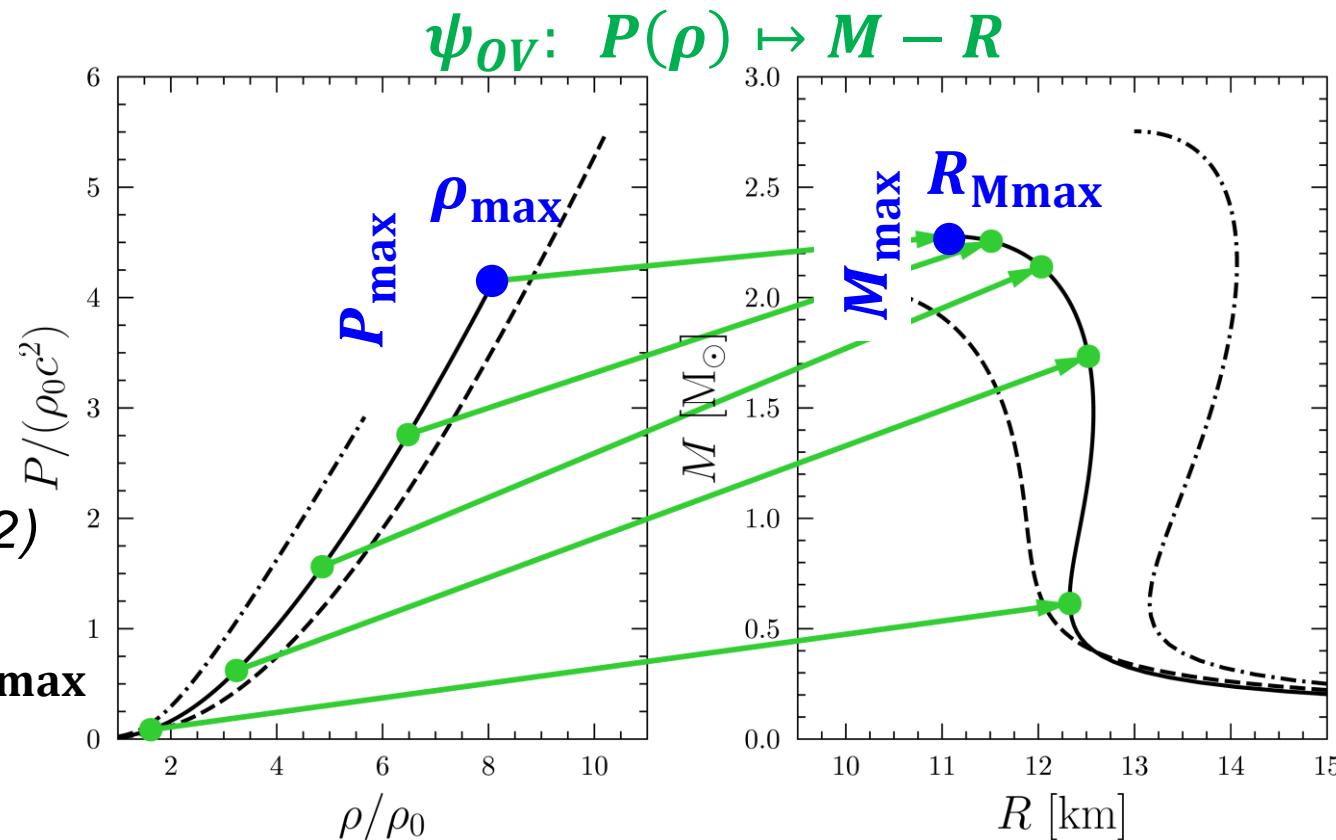
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 (Soma+ JCAP 2022)



- $\exists M_{\max} \leftrightarrow \rho_{\max}, P_{\max}$

# Dimension of EoS Manifold

Lindblom (2010): spectral representation

- 34 «realistic» EoS нреи нреиΛΣ
- universal fit нреиΛΣuds

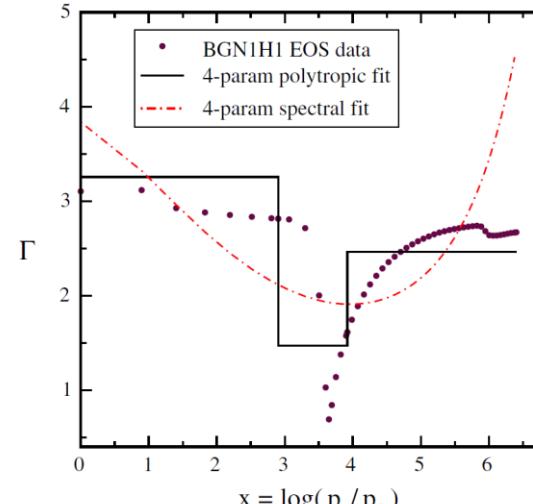
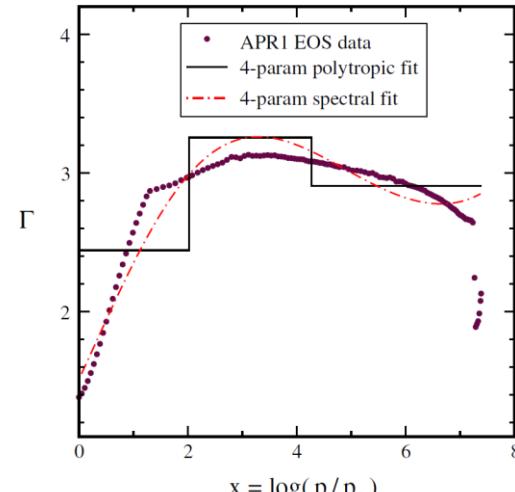
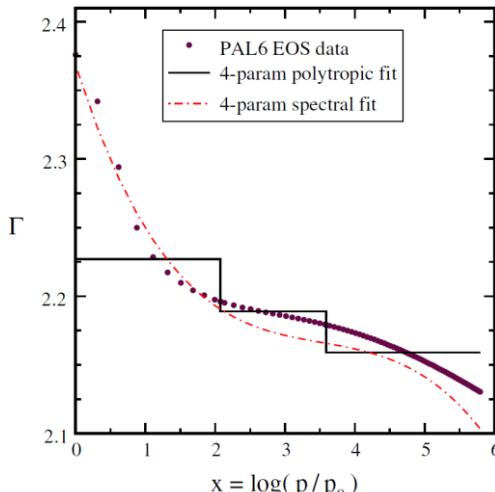
$$\ln \Gamma(P) = \underbrace{\gamma_0 + \gamma_1 \ln P}_{P - \rho} + \underbrace{\gamma_2 (\ln P)^2 + \gamma_3 (\ln P)^3 + \gamma_4 (\ln P)^4}_{M - R} + \dots$$

$rms \sim 3 - 8\%$

$rms \sim 1 - 5\%$

$rms \sim 0.5 - 4\%$

$$\Gamma = \frac{\rho c^2 + P}{\rho c^2} \frac{dP}{d(\rho c^2)}$$



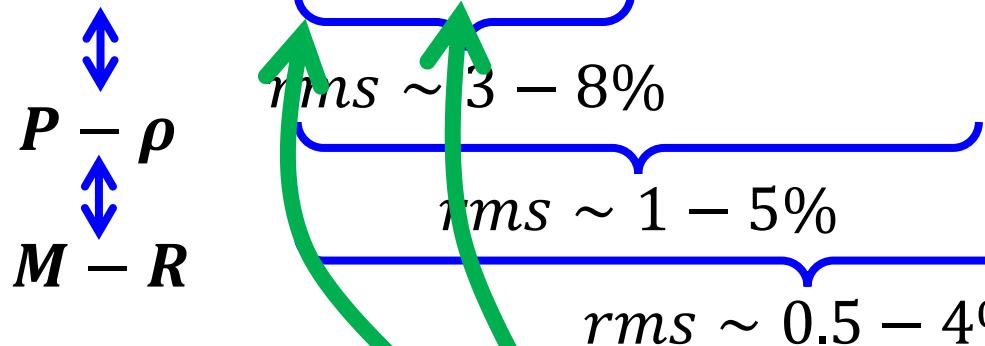
Lindblom (2010)

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$$\ln \Gamma(P) = \gamma_0 + \gamma_1 \ln P + \gamma_2 (\ln P)^2 + \gamma_3 (\ln P)^3 + \gamma_4 (\ln P)^4 + \dots$$



$$\Gamma = \frac{\rho c^2 + P}{\rho c^2} \frac{dP}{d(\rho c^2)}$$

Idea of  
this  
work

*handful  
parametrization:*

$$M_{\max}, R_{M\max} \leftrightarrow P_{\max}, \rho_{\max}$$

- self-similarity of  $M - R$  &  $P - \rho$  curves
- explicit  $\psi_{ov}^{-1}$

# Plan of the Further Talk

1. Describe the zoo of used EoSs

2. Build the Inverse OV mapping

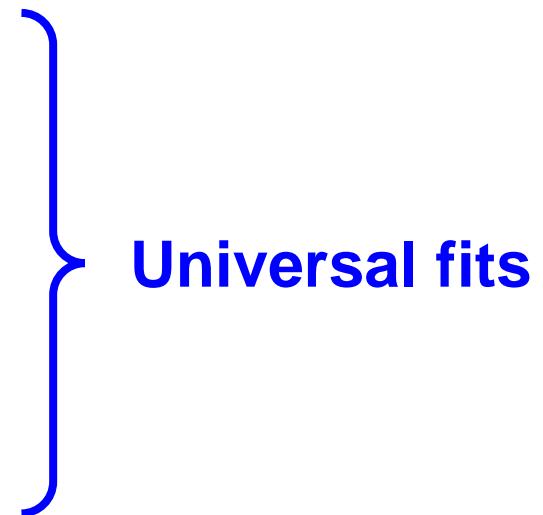
➤ **Step I:**  $P_{\max}, \rho_{\max} \leftrightarrow M_{\max}, R_{\max}$

➤ **Step II: self-similar curves**

$$P/P_{\max} - \rho/\rho_{\max}$$

➤ **Step III: self-similar curves**

$$M/M_{\max} - R/R_{\max}$$



3. Apply the Inverse OV to observations

# EoS Zoo

162  
models

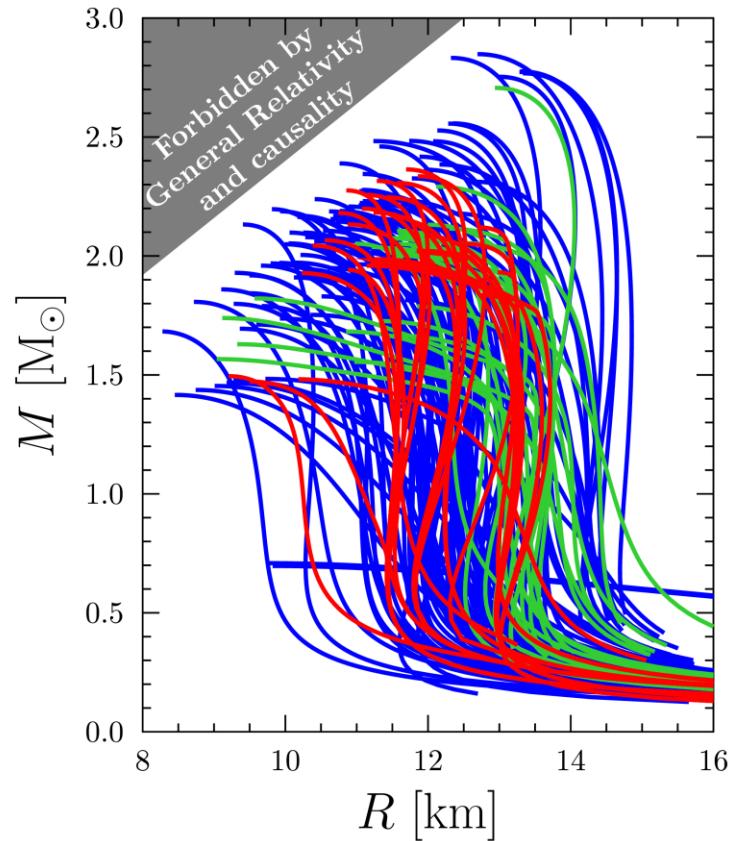
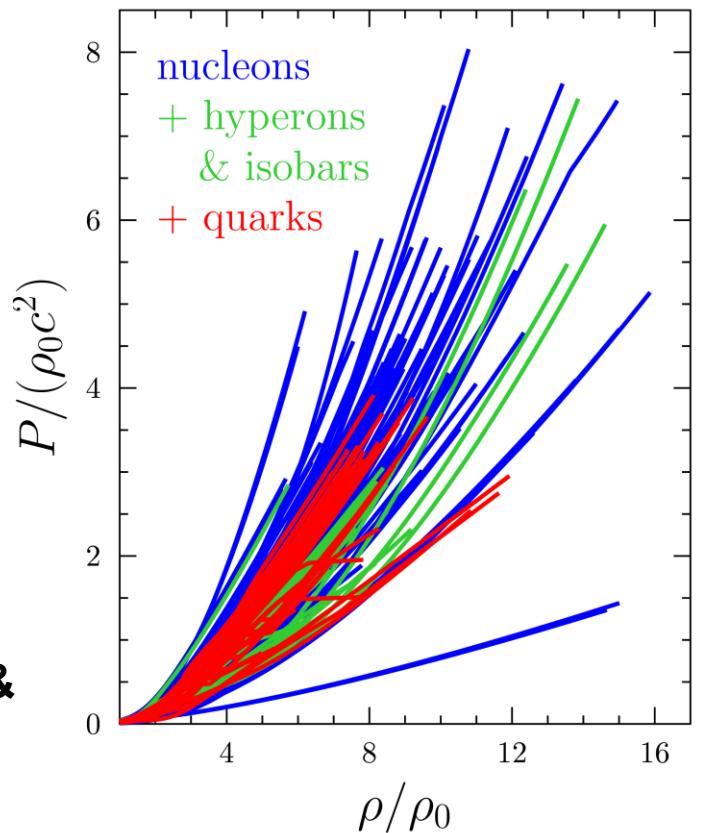
$\xrightarrow{\text{97 } npe\mu}$   $\xrightarrow{\text{32 } +\Lambda\Sigma\Xi\Delta}$   $\xrightarrow{\text{33 } +uds}$

- **CompOSE**  
[https://compose.  
obspm.fr/](https://compose.obspm.fr/)

- **Read+2009**  
 $\leftrightarrow$ Lindblom  
(2010)

- **Ozel & Freira**  
2016

- **Gusakov, Kantor & Haensel 2014,**  
**Fortin+2017,**  
**Ofengeim+2019,...**



...free  $npe$ , PAL, HHJ, variational, Skyrme, RMF, QMC, QHC,...

# EoS Zoo

162  
models

$\rightarrow 97 \text{ } npe\mu$

$\rightarrow 32 + \Lambda\Sigma\Xi\Delta$

$\rightarrow 33 + uds$

- **CompOSE**

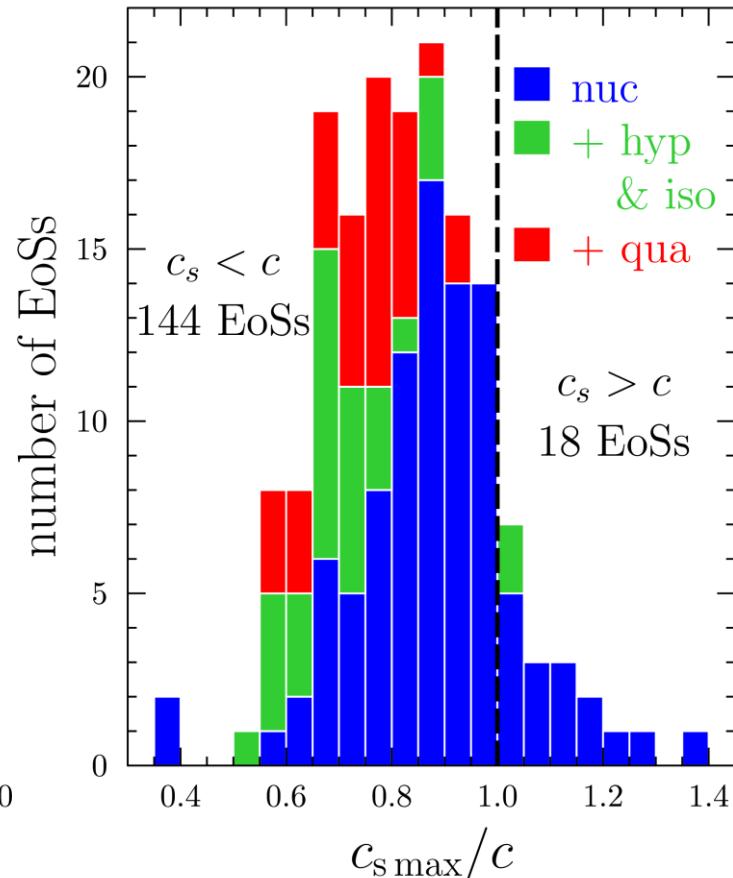
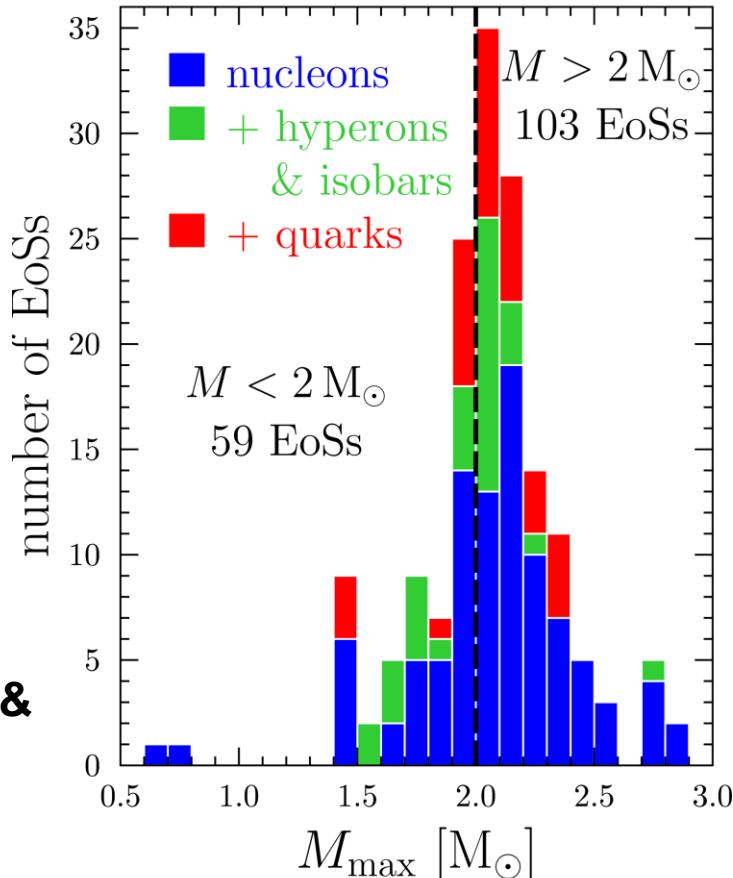
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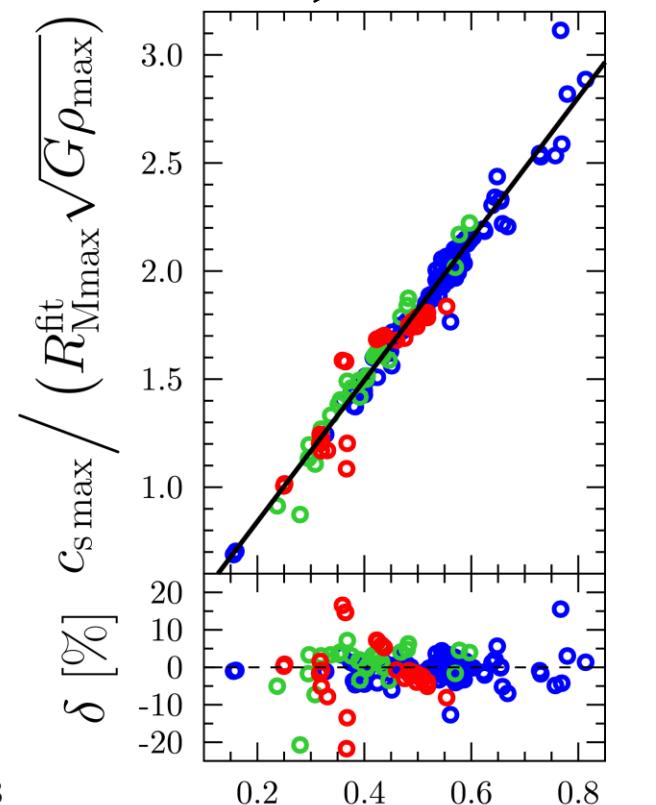
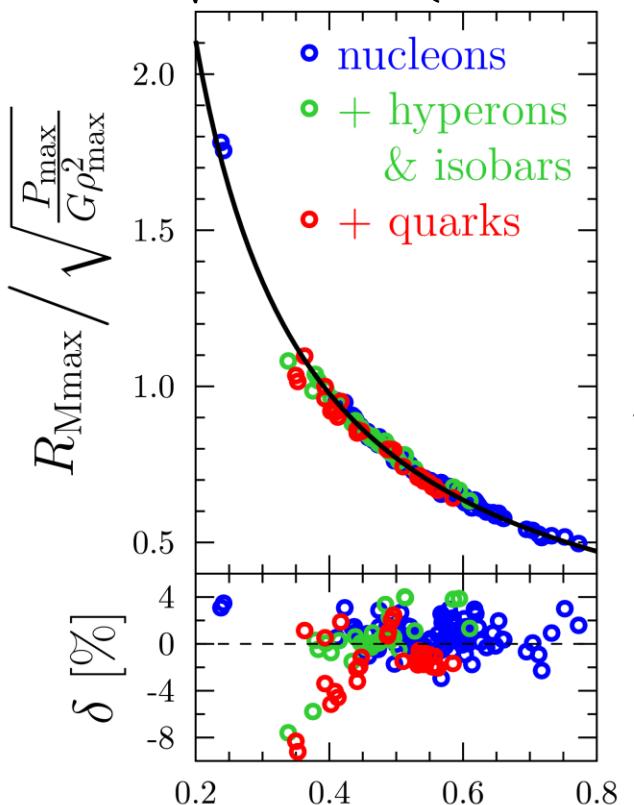
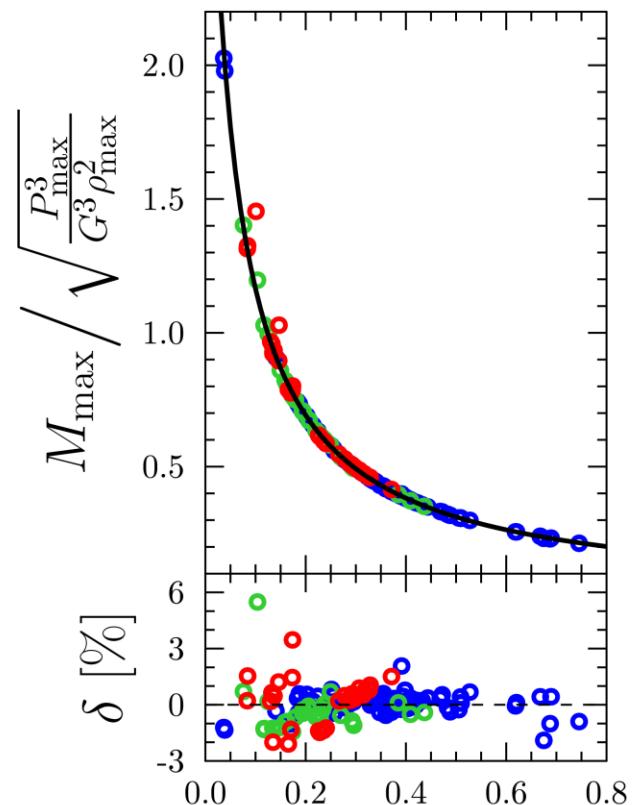
...free  $npe$ , PAL, HHJ, variational, Skyrme, RMF, QMC, QHC,...

# Step I: $M_{\max}, R_{\max} \leftrightarrow P_{\max}, \rho_{\max}$

$$M_{\max} = \sqrt{\frac{P_{\max}^3}{G^3 \rho_{\max}^2}} f_M \left( \frac{\rho_{\max}}{\rho_0}, \frac{P_{\max}}{\rho_0 c^2} \right)$$

$$R_{\max} = \sqrt{\frac{P_{\max}}{G \rho_{\max}^2}} f_R \left( \frac{\rho_{\max}}{\rho_0}, \frac{P_{\max}}{\rho_0 c^2} \right)$$

$$c_s \max = R_{M\max}^{\text{fit}} \sqrt{G \rho_{\max}} f_c \left( \rho_{\max}/\rho_0, P_{\max}/\rho_0 c^2 \right)$$

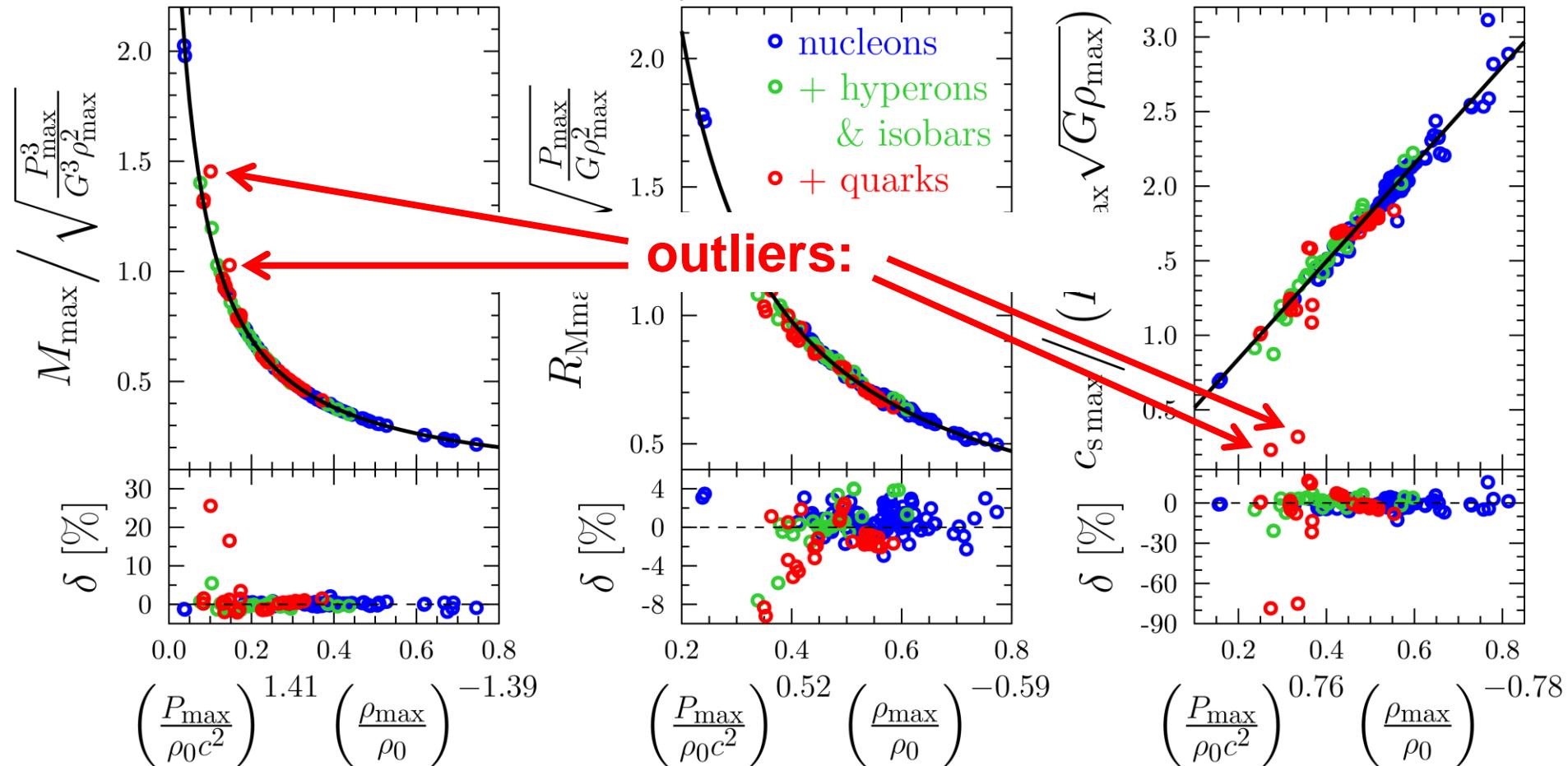


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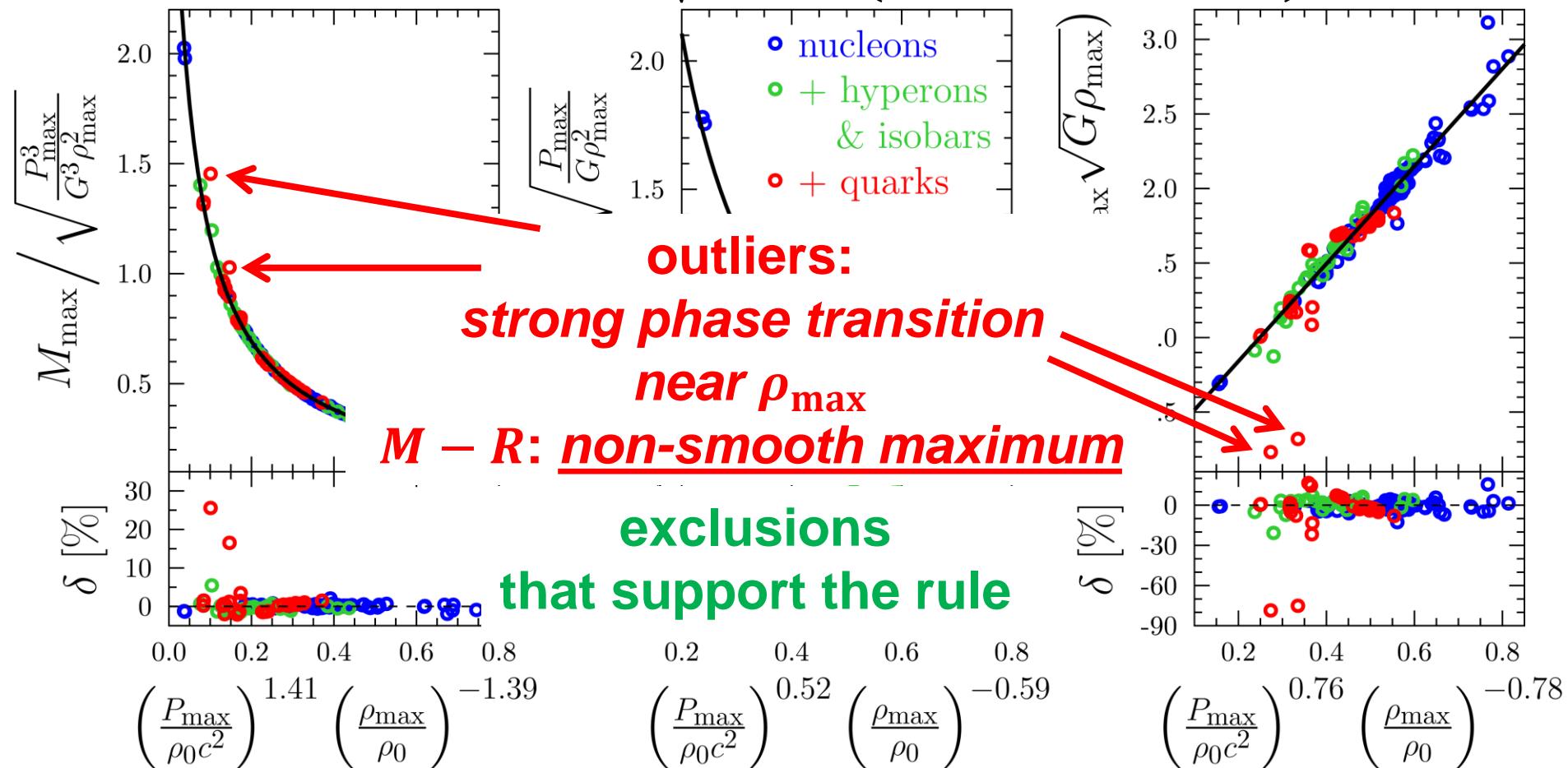


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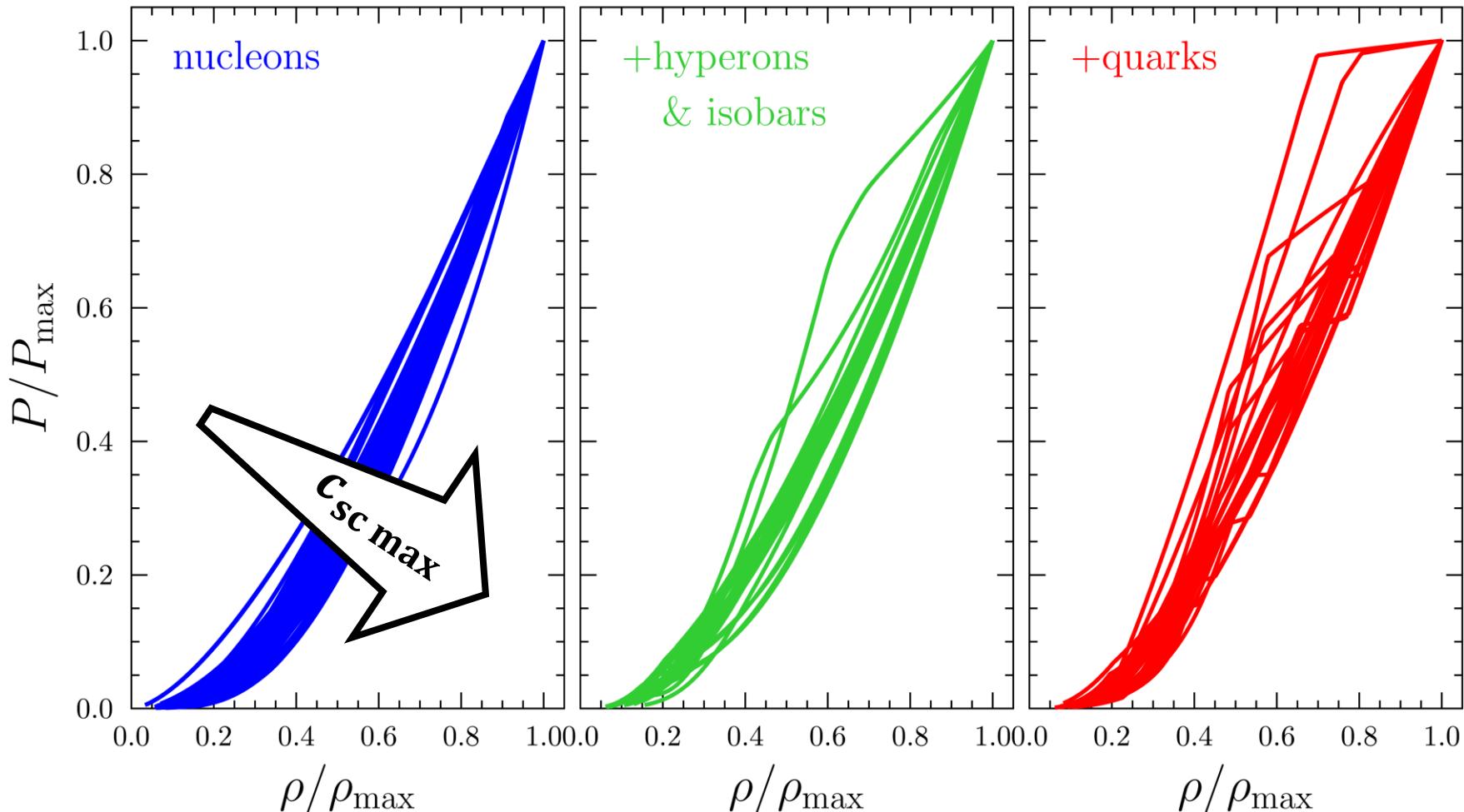
$$R_{\max} = \sqrt{\frac{P_{\max}}{G \rho_{\max}^2}} f_R \left( \frac{\rho_{\max}}{\rho_0}, \frac{P_{\max}}{\rho_0 c^2} \right)$$

$$c_s \max = R_{\max}^{\text{fit}} \sqrt{G \rho_{\max}} f_c \left( \rho_{\max}/\rho_0, P_{\max}/\rho_0 c^2 \right)$$



# Step II: Self-similar $P(\rho)$

$$\frac{P}{P_{\max}} \approx g\left(\frac{\rho}{\rho_{\max}}\right)$$



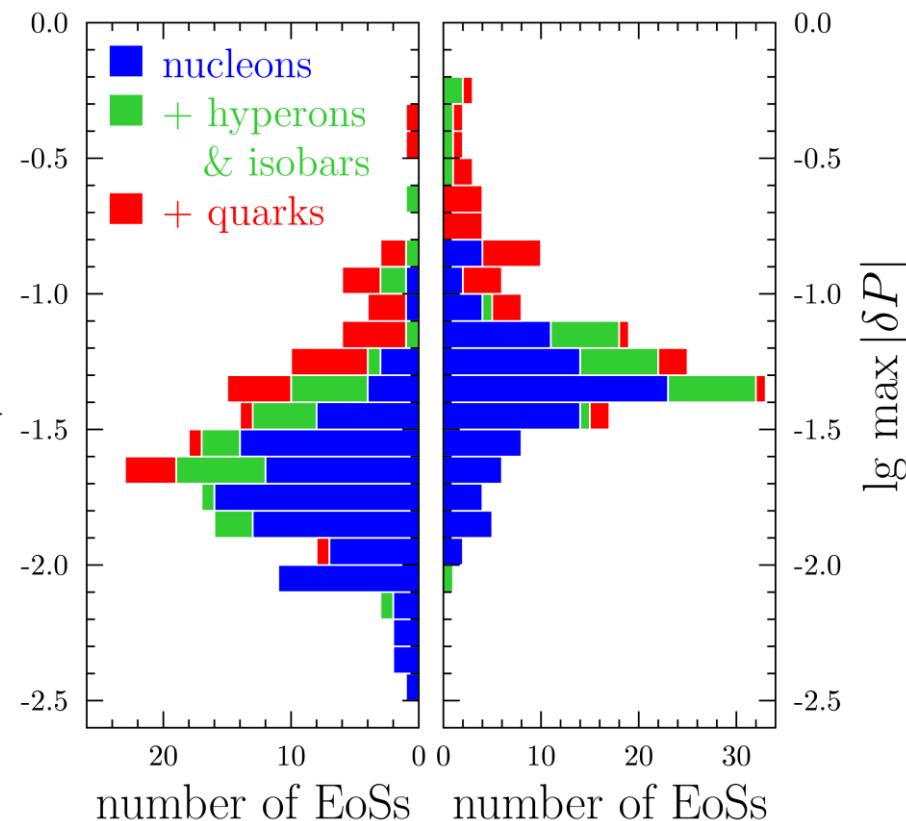
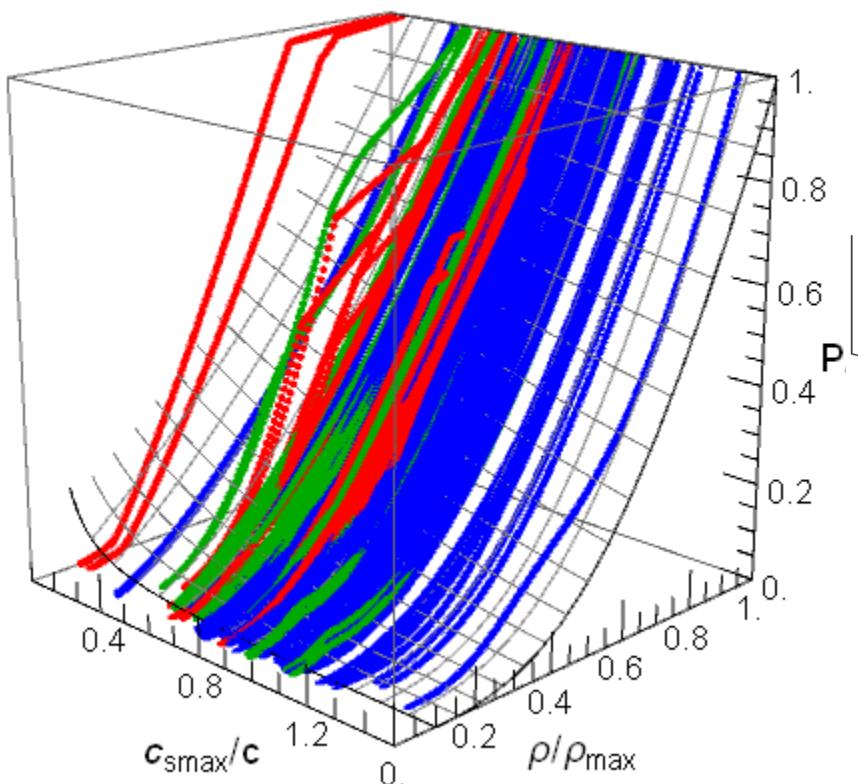
# Step II: Universal Fit $P(\rho)$

$$\frac{P}{P_{\max}} = g\left(\frac{\rho}{\rho_{\max}}; c_s \max, \gamma_{\max}\right)$$

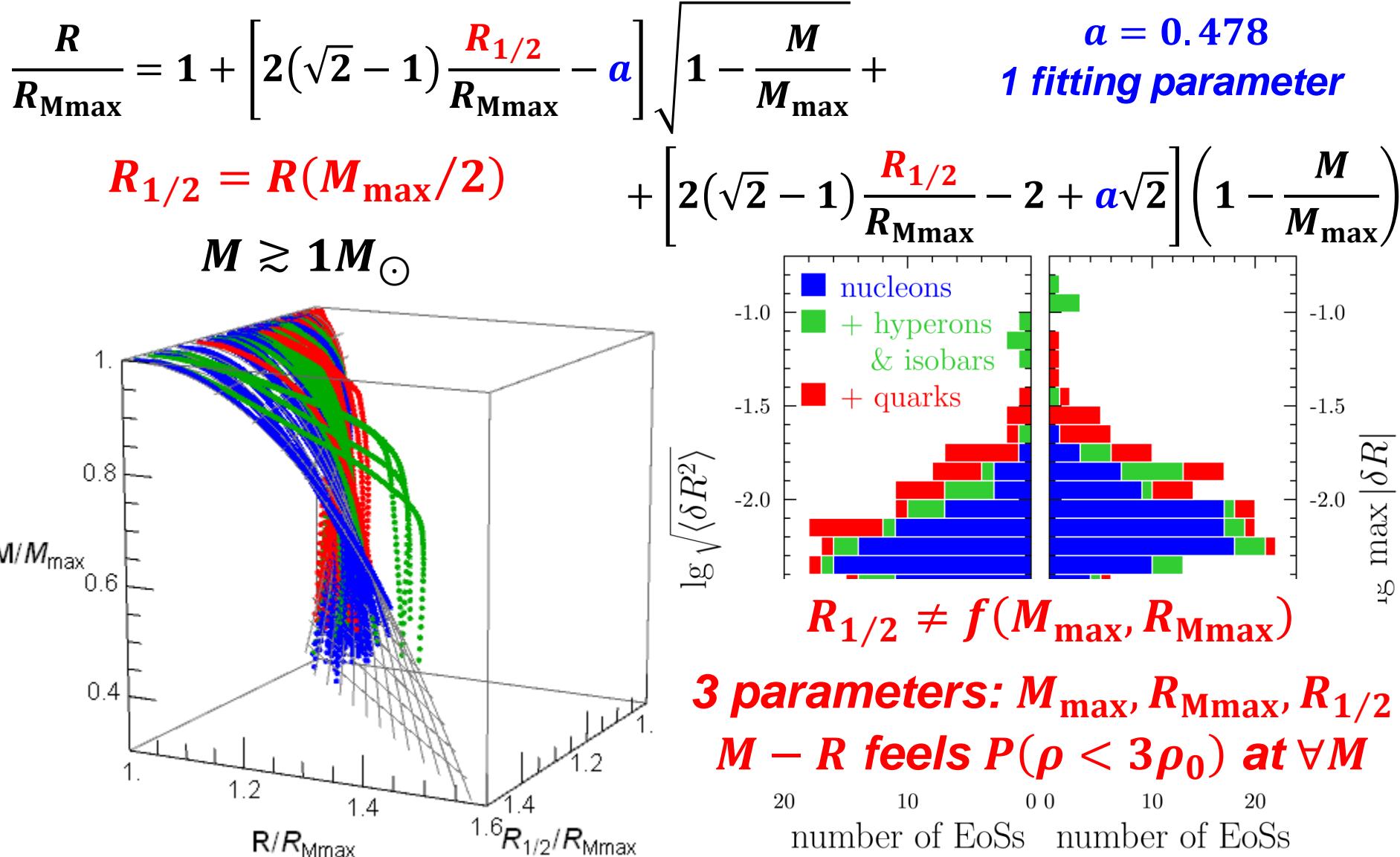
$c_s \max(P_{\max}, \rho_{\max})$

$$\rho \gtrsim 3\rho_0 \Leftrightarrow \text{center of } 1M_{\odot}$$

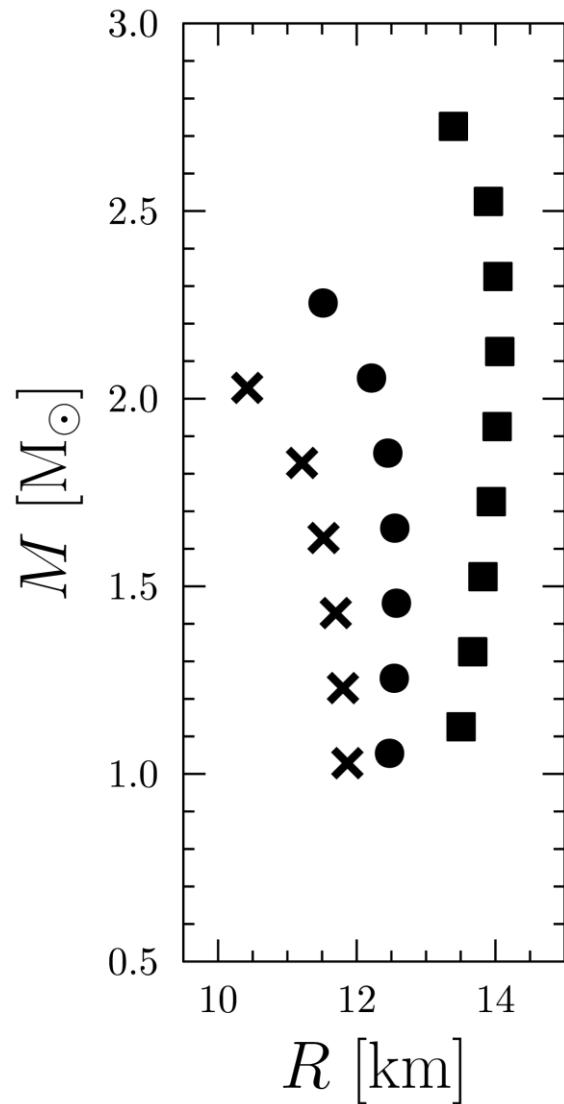
$$\gamma_{\max} = \frac{\rho_{\max}}{P_{\max}} c_s^2 \max(P_{\max}, \rho_{\max})$$



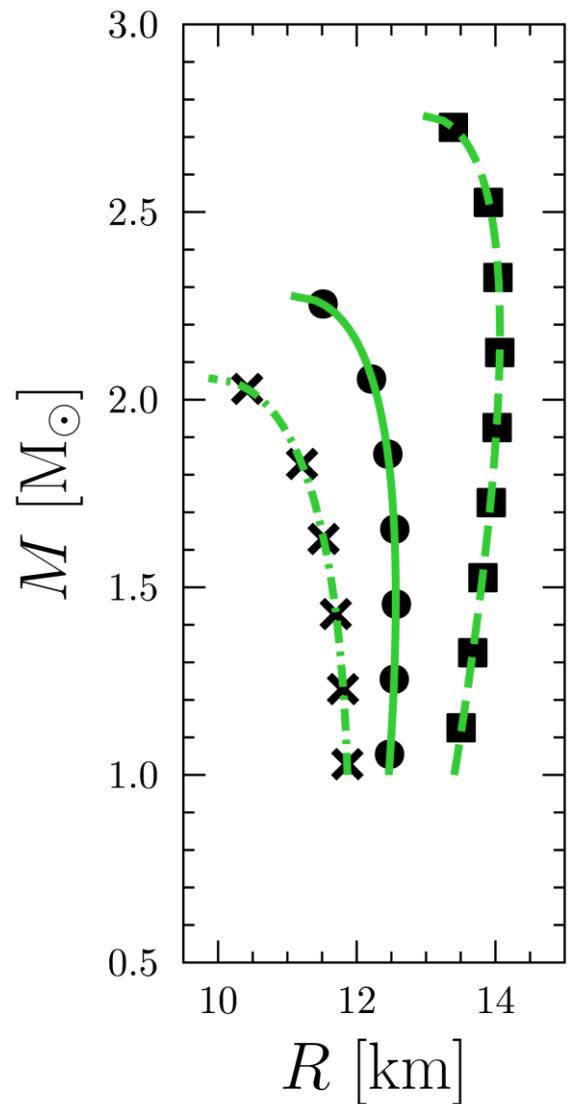
# Step III: Universal Fit $R(M)$



# Inverse Oppenheimer-Volkoff Mapping



# Inverse Oppenheimer-Volkoff Mapping



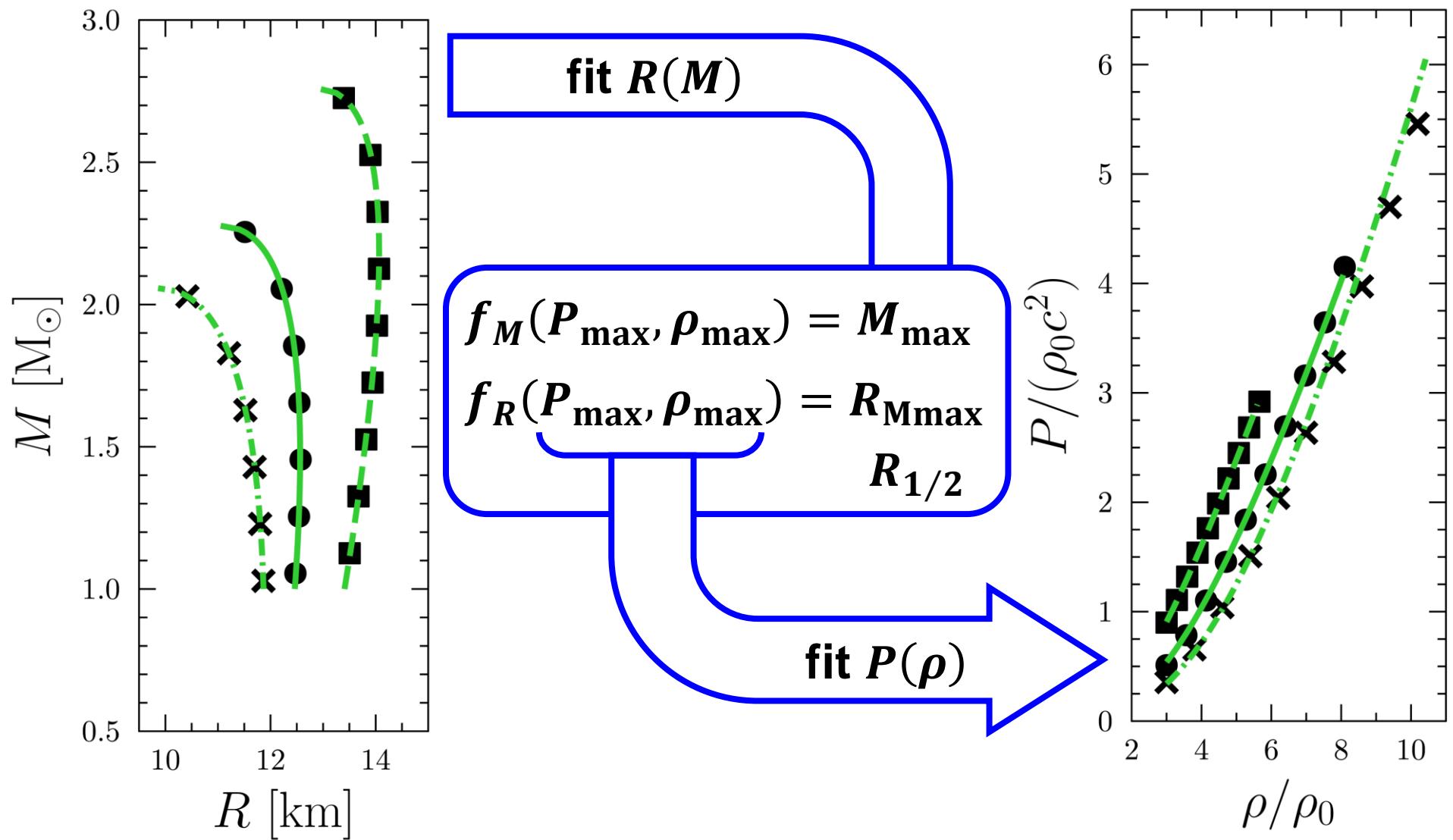
fit  $R(M)$

$$f_M(P_{\max}, \rho_{\max}) = M_{\max}$$

$$f_R(P_{\max}, \rho_{\max}) = R_{M_{\max}}$$

$$R_{1/2}$$

# Inverse Oppenheimer-Volkoff Mapping



# Analysis of Observations: Method

$M_{\max} >$  radiopulsars

$PSR\ J1614-2230$

$PSR\ J0348+0432$

Arzoumanian+'18; Fonseca+'21

**GWs:**  $\Lambda = f \left( \frac{2GM}{Rc^2} \right)$

Yagi & Younes'13, '16

$GW170817$

Abbot+'18

$M_i, R_i \in \left( \text{fit of NS spectrum} \right)_i$

$PSR\ J0740+6620$

Miller+'19; Riley+'19

$PSR\ J0030+0451$

Miller+'21; Riley+'21

Cas A Shternin+'23

fit  $R(M)$

$$c_s \max(P_{\max}, \rho_{\max}) < c$$

$$R_{1/2} > R_{M\max}$$

$$f_M(P_{\max}, \rho_{\max}) = M_{\max}$$

$$f_R(P_{\max}, \rho_{\max}) = R_{M\max}$$

$$R_{1/2}$$

fit  $P(\rho)$



# Analysis of Observations: Results

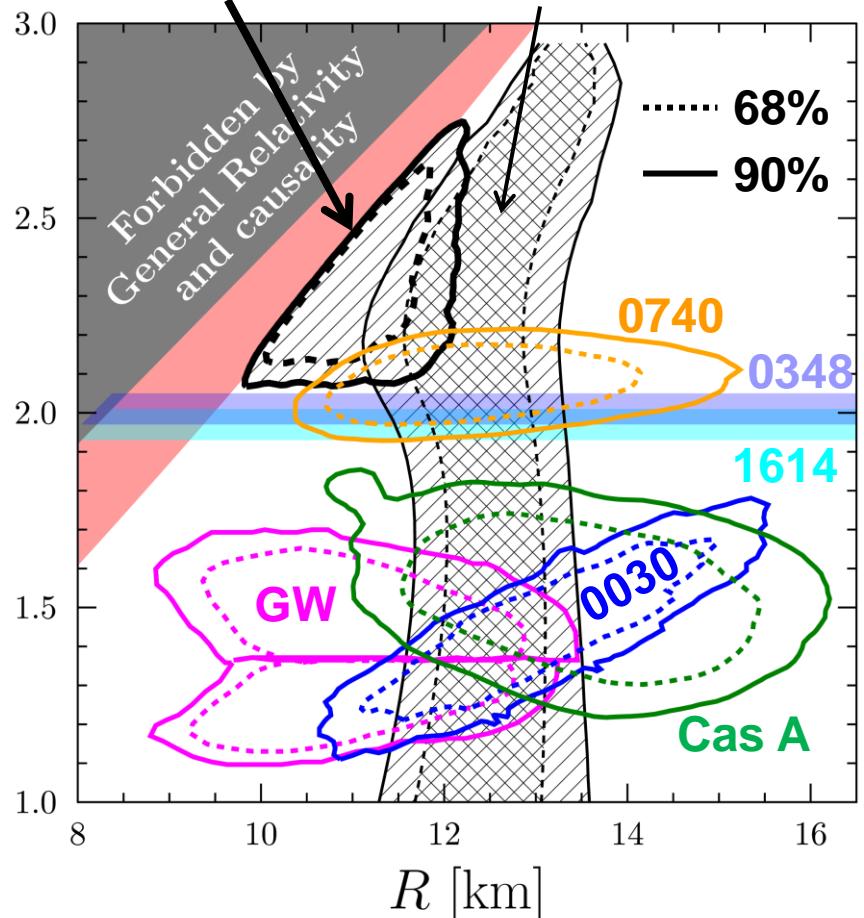
$$M_{\max} - R_{M\max}$$

$$R(M)$$

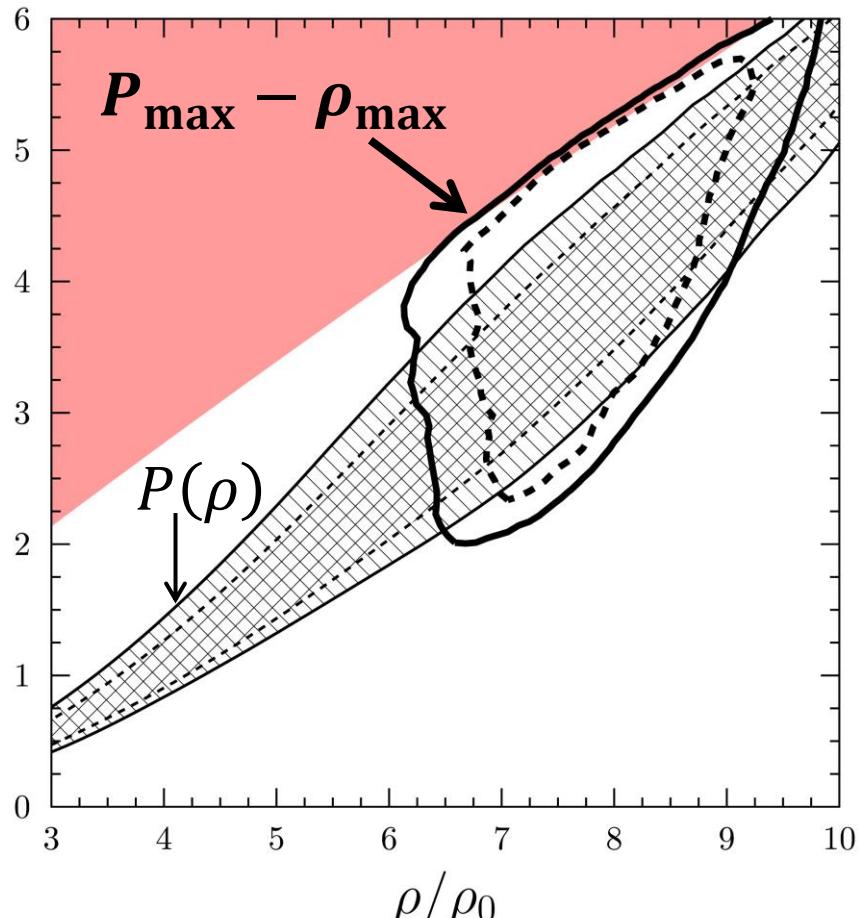


$$c_s \max(P_{\max}, \rho_{\max}) < c(+0.1c)$$

$$M [M_{\odot}]$$



$$P / (\rho_0 c^2)$$



$$M_{\max} = 2.29^{+0.15}_{-0.15} M_{\odot}$$

$$R_{\max} = 11.3^{+0.5}_{-0.6} \text{ km}$$

$$R_{1/2} = 12.4^{+0.7}_{-0.6} \text{ km}$$

$$P_{\max} = 4.1^{+1.0}_{-1.0} \rho_0 c^2$$

$$\rho_{\max} = 7.7^{+1.0}_{-0.8} \rho_0$$

# Add More Observations

$M_{\max} >$  radiopulsars

PSRs  $J1614-2230, J0348+0432$

+  
*spider*  
PSRs  $J0952-0607, J1311-3430,$   
 $J1653-0158, J1810+1744$   
Kandel&Romani'23

**GWs:**  $\Lambda = f \left( \frac{2GM}{Rc^2} \right)$  Yagi & Younes  
'13, '16

GW170817

$M_i, R_i \in \left( \begin{array}{l} \text{fit of NS} \\ \text{spectrum} \end{array} \right)_i$

PSRs  $J0740+6620, J0030+0451$

+ X-ray bursters Cas A

$4U 1702-429, 4U1724-307,$   
SAX J1810.8-2609

fit  $R(M)$

$c_s \max(P_{\max}, \rho_{\max}) < c$   
 $R_{1/2} > R_{M\max}$

$f_M(P_{\max}, \rho_{\max}) = M_{\max}$

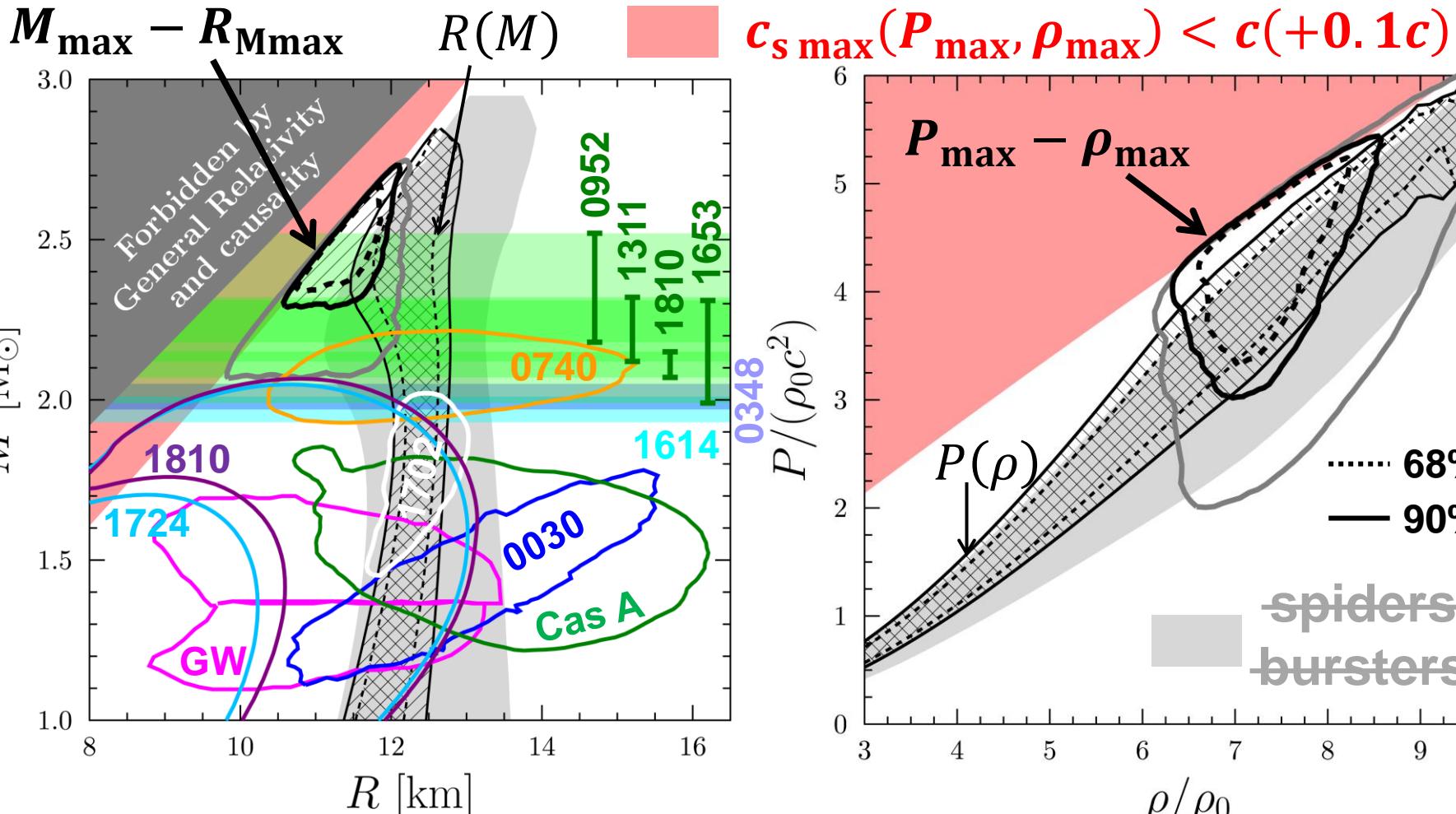
$f_R(P_{\max}, \rho_{\max}) = R_{M\max}$

$R_{1/2}$

fit  $P(\rho)$

?

# Results for More Observations



$$M_{\max} = 2.47^{+0.09}_{-0.12} M_\odot$$

$$R_{\max} = 11.5^{+0.3}_{-0.3} \text{ km}$$

$$R_{1/2} = 12.0^{+0.3}_{-0.3} \text{ km}$$

$$P_{\max} = 4.5^{+0.3}_{-0.8} \rho_0 c^2$$

$$\rho_{\max} = 7.3^{+0.5}_{-0.5} \rho_0$$

# Discussion

- Physics or antropology?
- Extend to  $\rho < 3\rho_0$ ?
  - *Lattimer & Prakash 2001:  $R_{1.4} \propto P^{1/4}(2n_0)$*
- Extend to  $M < 1M_\odot$ ?
  - *Sufficient dimension of  $M - R$  curves family?*
- Account for rotation?
  - *Talk by A. Konstantinou*

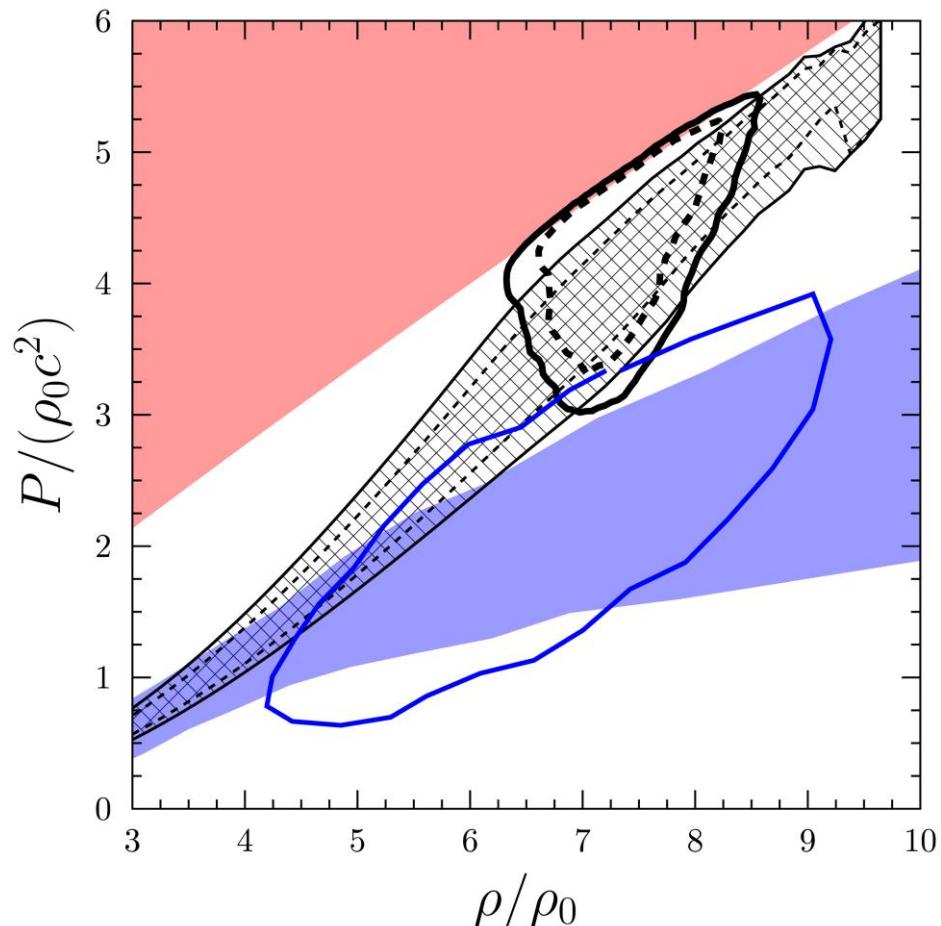
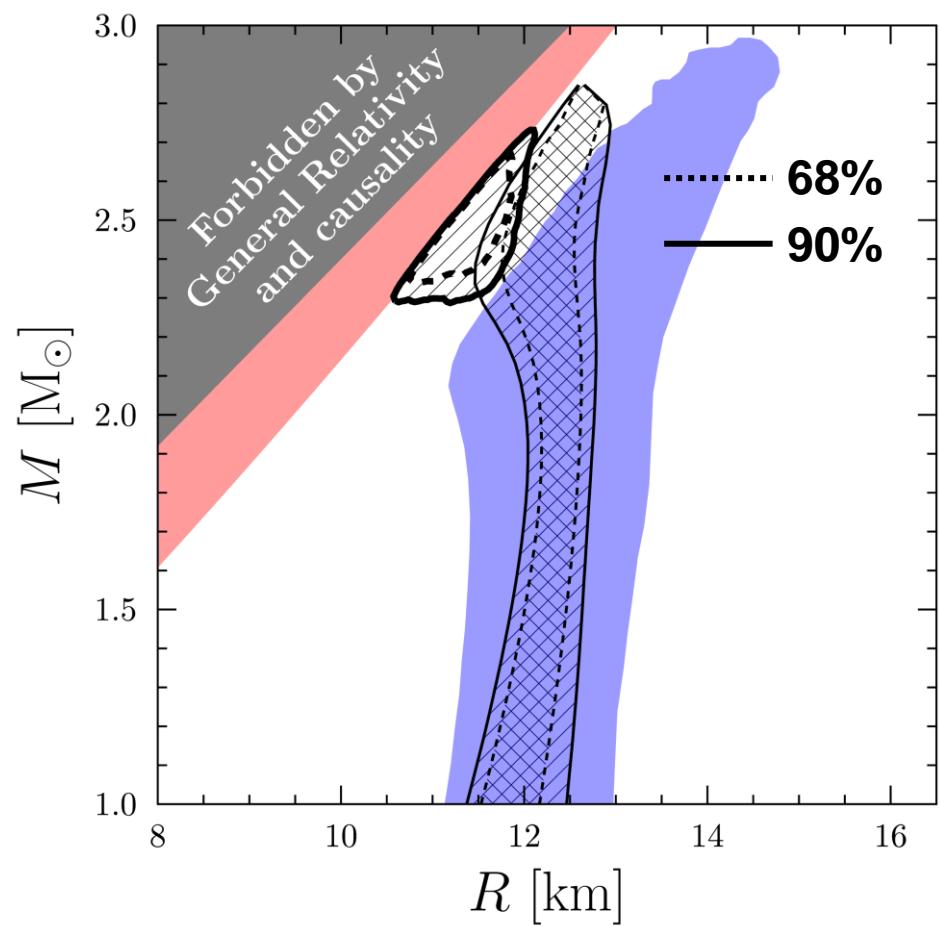
# Conclusion

- Maximum-mass NS – handful universal scale of hydrostatic properties of NSs
- Using this scale we
  - provide universal fits for  $P - \rho$  and  $M - R$
  - derived explicit (semi)analytic inverse Oppenheimer-Volkoff mapping
- This inverse OV mapping – new handful tool to gain properties of superdense matter from NS observations

# Thank you!

***More equations of state are welcome***

# Comparison with Other Works



90% Jiang, Ecker & Rezzolla 2023