General Motivation
Observational constraints
Tolman-Oppenheimer-Volkoff equations
Bayesian Analysis
Results Conclusions
Motivation

# Bayesian Analysis of Hybrid EoS based on Astrophysical Observational Data

## Alexander Ayriyan<sup>1</sup>

D. Alvares<sup>2,3</sup>, D. Blaschke<sup>3,4</sup> and H. Grigorian<sup>1,5</sup>

<sup>1</sup>Laboratory of Information Technologies, JINR
 <sup>2</sup>Instituto de Física, Universidad Autónoma de San Luis Potosí
 <sup>3</sup>Bogoliubov Laboratory for Theoretical Physics, JINR
 <sup>4</sup>Institute for Theoretical Physics, University of Wroclaw
 <sup>5</sup>Department of Theoretical Physics, Yerevan State University

August 25, 2014



## Qualification and Classification of EoS

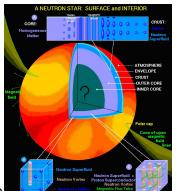
- Estimation of different models of EoS from observational constraints
- Applying Bayesian Analysis for the estimation
- Finding suggestions for observation which could be most selective for the models of EoS





General Motivation
Observational constraints
Tolman-Oppenheimer-Volkoff equations
Bayesian Analysis
Results Conclusions
Motivation

## Neutron Star Structure



Credit: Dany Page

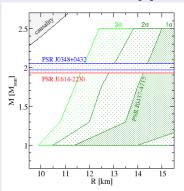




## **Observational Constraints**

#### Mass and Radius Constraints

Radius and maximum mass constraints are given from PSR J0437-4715 [1] and PSR J0348+0432 [2] correspondingly.



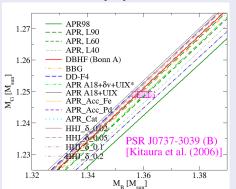




## **Observational Constraints**

#### **Gravitational Binding Energy Constraint**

A constraint on the gravitational binding energy is taken from the neutron star B in the binary system J0737-3039 (B) [3].



formation :hnologies



## **Observational Constraints**

## Three Statistically Independent Constraints

- A radius constraint from the nearest millisecond pulsar PSR J0437-4715 [1].
- A maximum mass constraint from PSR J0348+0432 [2].
- A constraint on the gravitational binding energy from the neutron star B in the binary system PSR J0737-3039
   (B) [3].





# Tolman-Oppenheimer-Volkoff equations

#### **TOV** equations

$$\begin{cases} \frac{dm(r)}{dr} = C_1 \epsilon r^2 \\ \frac{dm_B(r)}{dr} = C_1 n_B m_N \frac{r^2}{(1 - 2C_2 m/r)} \\ \frac{dp(\epsilon, r)}{dr} = -C_2 \frac{(\epsilon + p)(m + C_1 p r^3)}{r(r - 2C_2 m)} \end{cases}$$
(1)

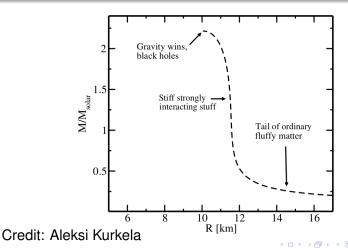
#### Constants

$$C_1 = 1.11269 \cdot 10^{-5} \frac{M_{\odot}}{\text{km}^3} \frac{\text{fm}^3}{\text{MeV}}$$
  $C_2 = 1.4766 \frac{\text{km}}{M_{\odot}}$ 

formation :hnologies

General Motivation Observational constraints Tolman-Oppenheimer-Volkoff equations Bayesian Analysis Results Conclusions Motivation

## Mass-Radius plot



# Tolman-Oppenheimer-Volkoff equations

#### **TOV** equations

$$\begin{cases}
\frac{dp}{d\epsilon} = F(m, dp/d\epsilon, r, \epsilon) \\
\frac{dm}{d\epsilon} = C_1 \epsilon r^2 F(m, dp/d\epsilon, r, \epsilon) \\
\frac{dm_B}{d\epsilon} = C_1 n_B m_N \frac{r^2}{(1 - 2C_2 m/r)} F(m, dp/d\epsilon, r, \epsilon)
\end{cases} (2)$$

$$F(m, dp/d\epsilon, r, \epsilon) = \frac{dp/d\epsilon}{-C_2 \frac{(\epsilon + p)(m + C_1 pr^3)}{r(r - 2C_2 m)}}$$
(3)

formation chnologies



## **EoS Parametrization**

#### Hybrid EoS

$$p(\epsilon) = p^{l}(\epsilon) \Theta(\epsilon_{c} - \epsilon) + p^{l}(\epsilon_{c}) \Theta(\epsilon - \epsilon_{c}) \Theta(\epsilon_{c} - \epsilon + \Delta \epsilon) + p^{ll}(\epsilon) \Theta(\epsilon - \epsilon_{c} - \Delta \epsilon),$$

where  $p^{l}(\epsilon)$  is given by a pure hadronic EoS (here well known model of APR), and  $p^{ll}(\epsilon)$  represents the high density nuclear matter [4] used here as quark matter given in the bag-like form.

#### Bag-Like Form of QM EoS

$$p^{II}(\epsilon) = c_{QM}^2 \epsilon - B,$$

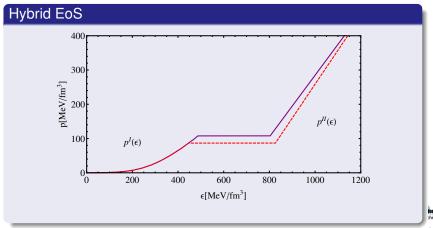
where  $c_{QM}^2$  is the squared speed of sound in quark matter and B is the bag constant.



General Motivation
Observational constraints
Tolman-Oppenheimer-Volkoff equations
Bayesian Analysis
Results Conclusions
Motivation

# EoS Models Formulation of the Problem Calculation of Probabilities

## **EoS** Parametrization



## **EoS** Parametrization

#### Hybrid EoS Pareameters

#### Vector of Parameters

For the BA, we have to sample the above defined parameter space and to that end we introduce a vector of the parameter values:

values: 
$$\pi_i = \overrightarrow{\pi} \left( \epsilon_{\mathcal{C}}(k), \gamma(I), c_{\mathrm{QM}}^2(m) \right),$$
  $i = 1 \dots N$  (here  $N = \prod_{q=1}^3 N_q$ ) and  $i = N_1 \times N_2 \times k + N_2 \times I + m$ 

fermation :hnologies

## Qualification of EoS Set from Observation

#### Goal

To find the set  $\pi_i$  corresponding to an EoS and thus a sequence of configurations which contains the most probable one based on the given constraints using BA (calculate of *a posteriori* probabilities of  $\pi_i$ ).

#### Unification of a priori probabilities

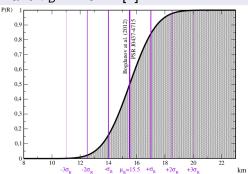
$$P(\pi_i) = 1$$
 for  $\forall i$ .





#### Probability of Corresponding to Radius Constraint for $\pi_i$

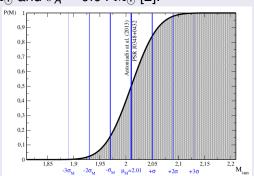
 $P(E_B | \pi_i) = \Phi(R_i, \mu_B, \sigma_B)$ , here  $R_i$  is max radius given by  $\pi_i$ .  $\mu_B = 15.5$  km and  $\sigma_B = 1.5$  km [1].





#### Probability of Corresponding to Mass Constraint for $\pi_i$

 $P(E_A | \pi_i) = \Phi(M_i, \mu_A, \sigma_A)$ , here  $M_i$  is max mass given by  $\pi_i$ .  $\mu_A = 2.01 \text{ M}_{\odot}$  and  $\sigma_A = 0.04 \text{ M}_{\odot}$  [2].





## Probability of Corresponding to $M - M_B$ Constraint for $\pi_i$

We need to estimate the probability for the closeness of a theoretical point  $M_i = (M_i, M_{Bi})$  to the observed point  $\mu_K = (\mu_G, \mu_B)$ . The required probability can be calculated using the following formula

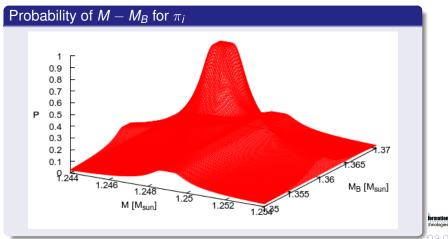
$$P(E_K | \pi_i) = [\Phi(\xi_G) - \Phi(-\xi_G)] \cdot [\Phi(\xi_B) - \Phi(-\xi_B)],$$

where  $\Phi(x) = \Phi(x, 0, 1)$ ,  $\xi_G = \sigma_{M_G}/d_{M_G}$  and  $\xi_B = \sigma_{M_B}/d_{M_B}$ , with  $d_{M_G}$  and  $d_{M_B}$  being the absolute values of components of the vector  $\mathbf{d}_i = \mu - \mathbf{M}_i$ , where  $\mu_{\mathbf{B}} = (\mu_G, \mu_B)^T$  is given in

ormation hnologies



## Calculation of Probabilities



EoS Models Formulation of the Problem Calculation of Probabilities

## Calculation of Probabilities

#### Probability of All Constraints for $\pi_i$

Taking to the account assumption that these measurements are independent on each other we can calculate complete conditional probability:

$$P(E|\pi_i) = P(E_A|\pi_i) \times P(E_B|\pi_i) \times P(E_K|\pi_i)$$

#### Calculation of a posteriori Probabilities of $\pi_i$

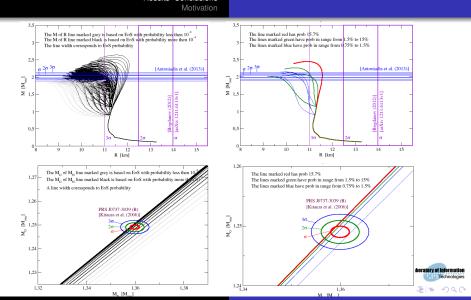
Now, we can calculate probability of  $\pi_i$  using Bayes' theorem:

$$P(\pi_{i}|E) = \frac{P(E|\pi_{i})P(\pi_{i})}{\sum\limits_{j=0}^{N-1}P(E|\pi_{j})P(\pi_{j})}$$

thnologies

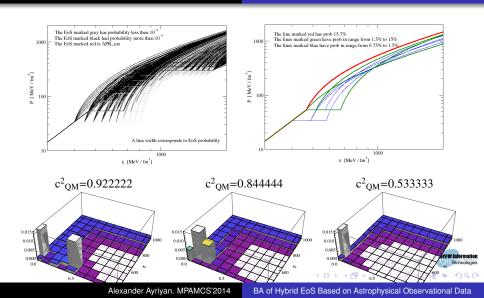
General Motivation
Observational constraints
Tolman–Oppenheimer–Volkoff equation
Bayesian Analysis
Results Conclusions

M-R and  $M_g$ - $M_B$  plots EoS plots



General Motivation
Observational constraints
Tolman-Oppenheimer-Volkoff equations
Bayesian Analysis
Results Conclusions
Motivation

M–R and  $M_g$ – $M_B$  plots EoS plots



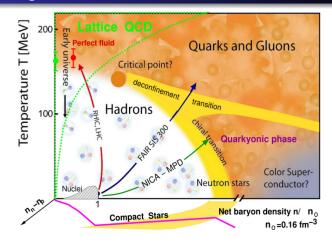
## Conclusions

- The most probable set of parameters resulting from the Bayesian Analysis point out to a quite stiff EoS with a smooth phase transition.
- Less probable configurations have jump in phase transition. Most of these EoS are pretty much stiff as well.
- The 7 most probable EoS do not allow a "third family".





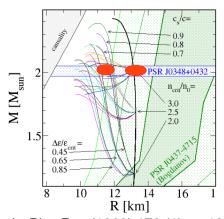
# Phase Diagram







## Fake measurements



Ulrich H. Gerlach. PhysRev (1968) 172 (1), p. 1325-1330.

#### References I

- S. Bogdanov. Astrophys. J. 762, 96 (2013)
- J. Antoniadis et al. Science 340, 6131 (2013)
- F.S. Kitaura et al. Astron. & Astrophys. 450, 345 (2006)
- M. Alford, S. Han and M. Prakash. Phys. Rev. D 88, 083013 (2013)
- M.G. Alford, S. Han and M. Prakash. Phys. Rev. D 88, 083013 (2013)





## References II









#### In the end, there can be only one.

- Duncan MacLeod

Thanks for your attention!



