

*From chiral EFT to perturbative QCD:*  
a Bayesian model mixing approach to the  
dense matter equation of state

Alexandra C. Semposki

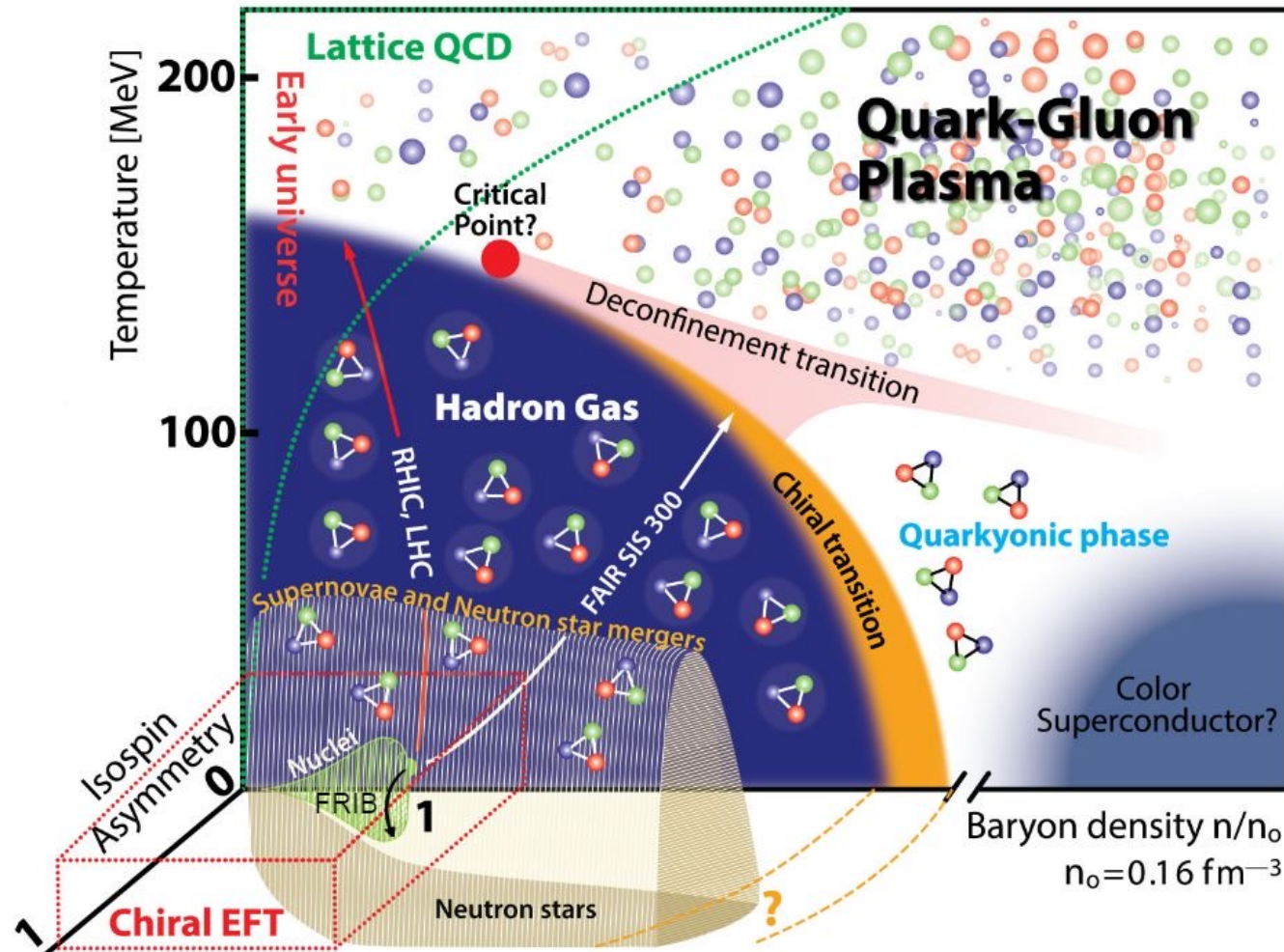
*in collaboration with:* C. Drischler, R. J. Furnstahl, J. A. Melendez, D. R. Phillips

arXiv:2404.06323v2



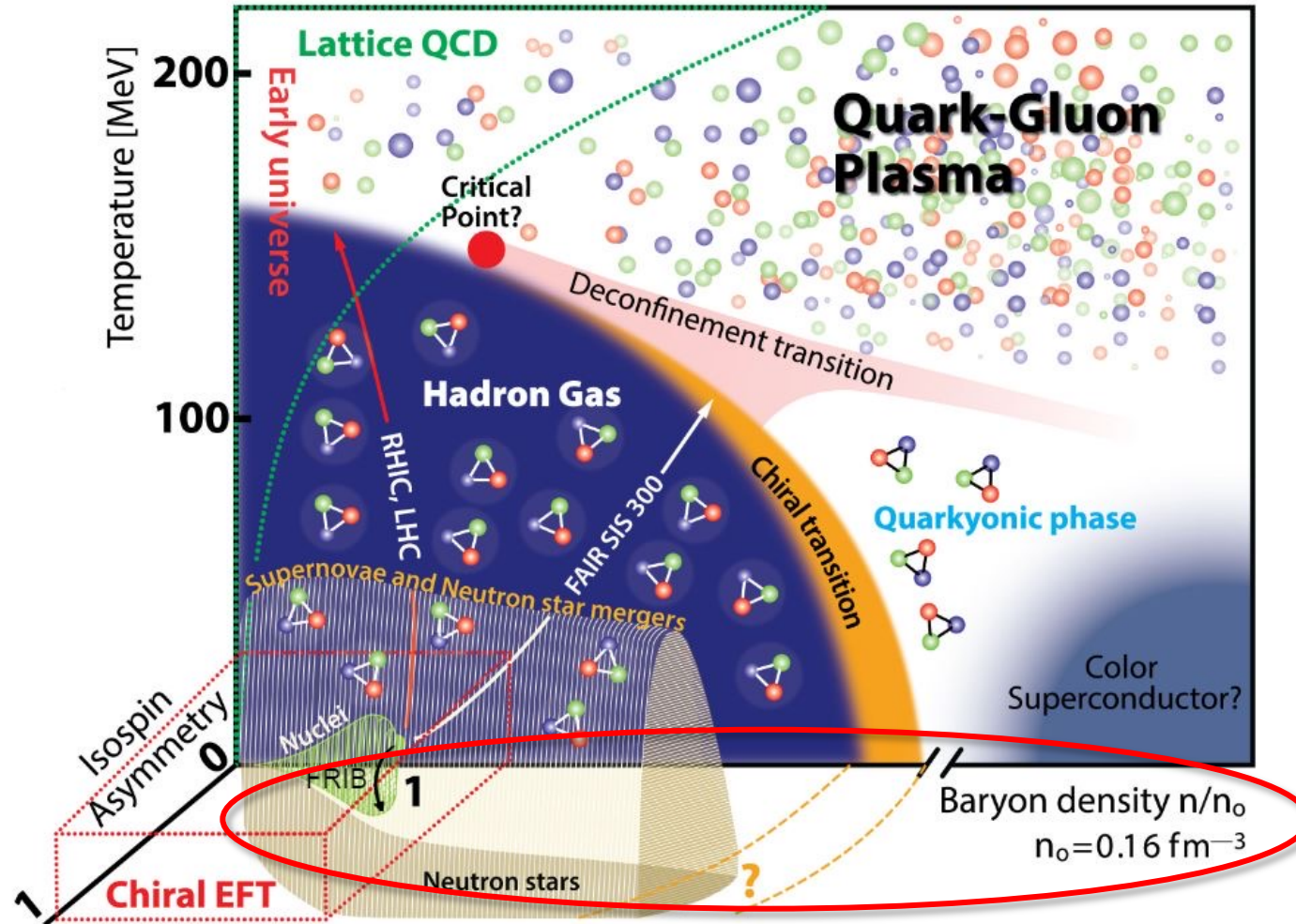
# Motivation: a unified EOS

Drischler et al. (2021)



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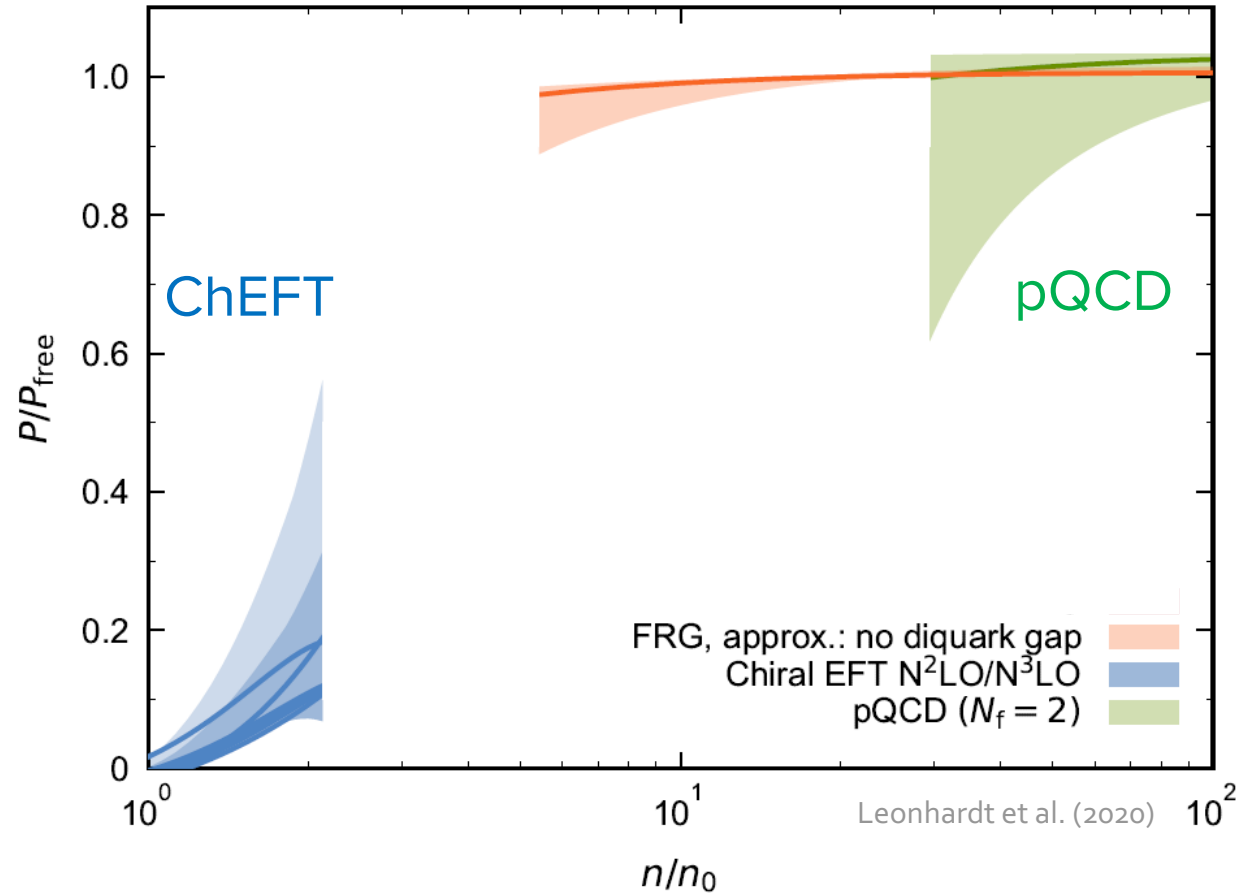
We are here

# Motivation: a unified EOS

How to develop an overall model from several individual models for the EOS?

[Symmetric Nuclear Matter]

Our goal: improve current estimates & constrain the gap using UQ

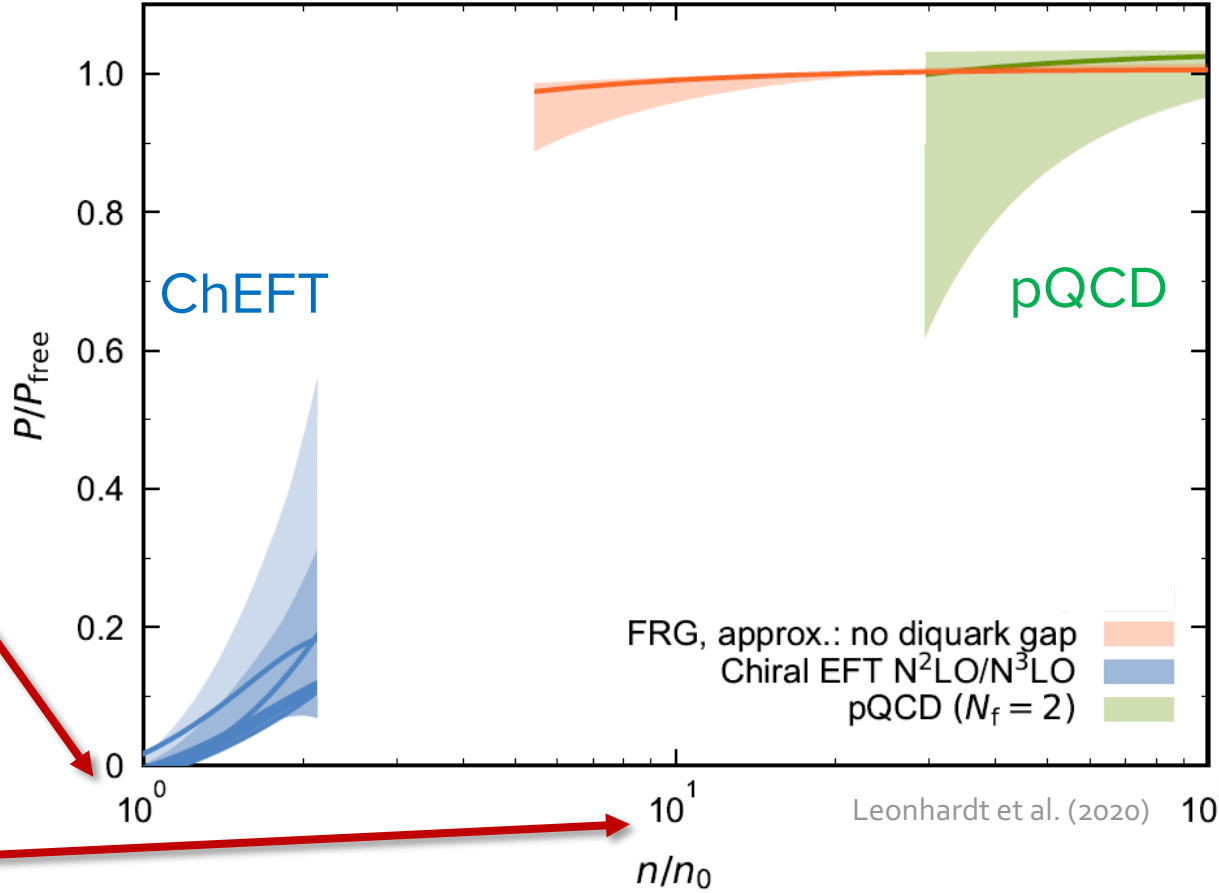
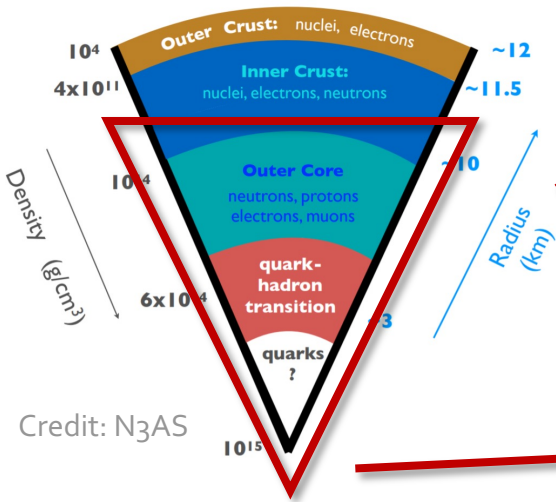


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Low densities can be described by  $\chi$ EFT, constraints from heavy-ion collisions



Credit: N<sub>3</sub>AS

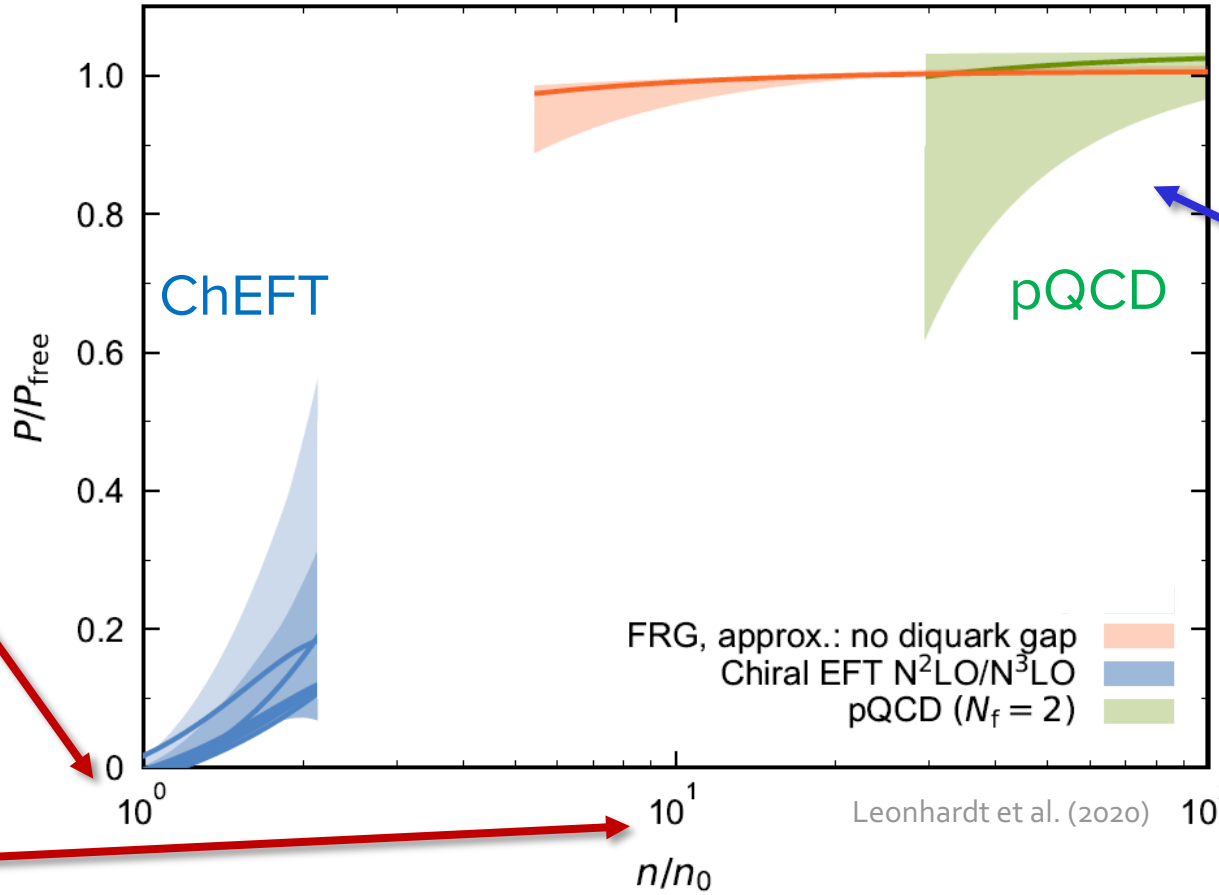
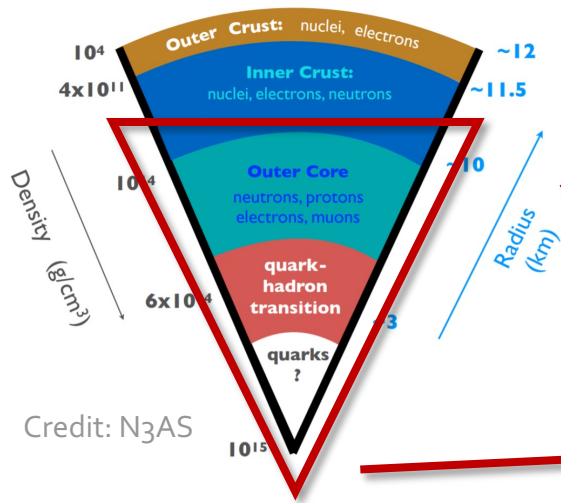
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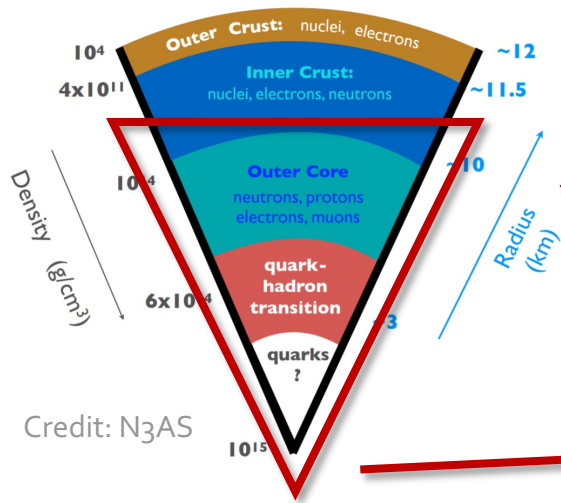
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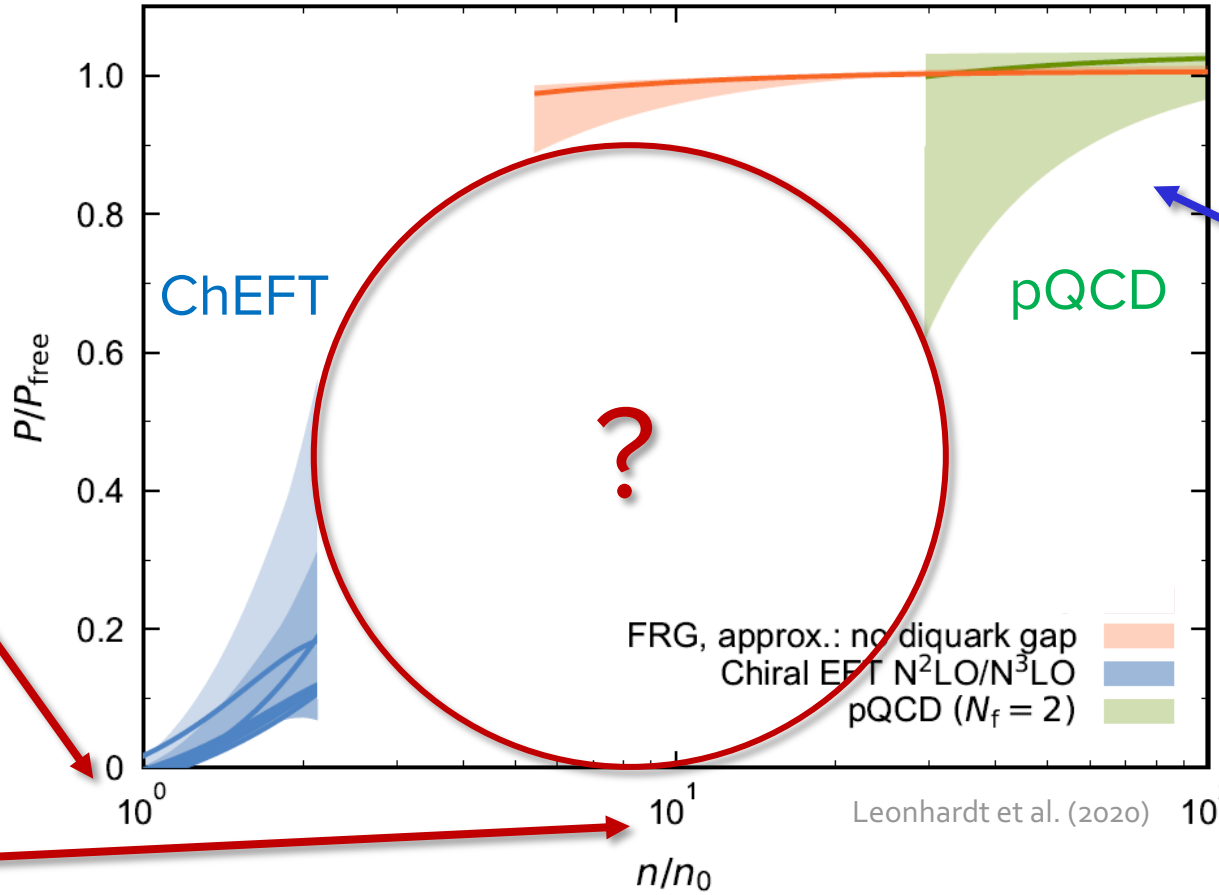
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Leonhardt et al. (2020)

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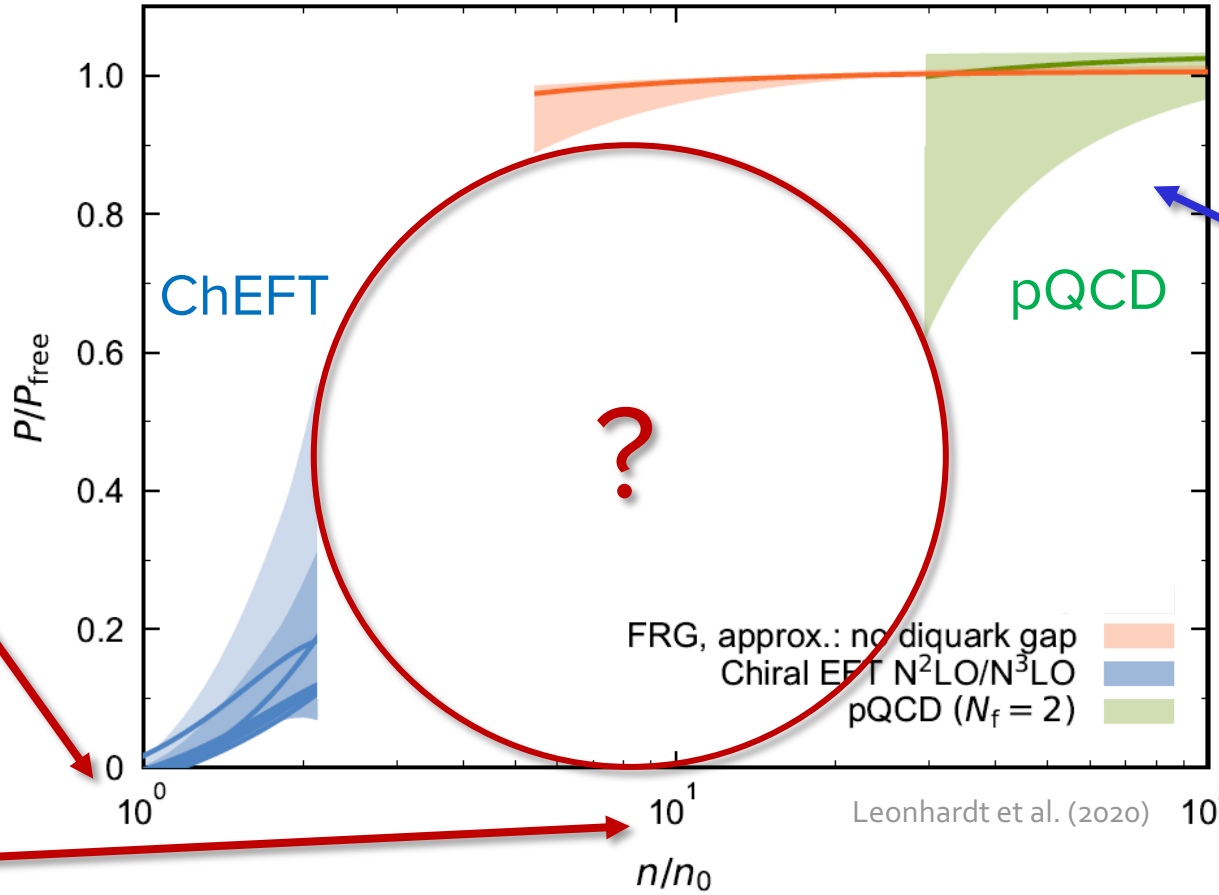
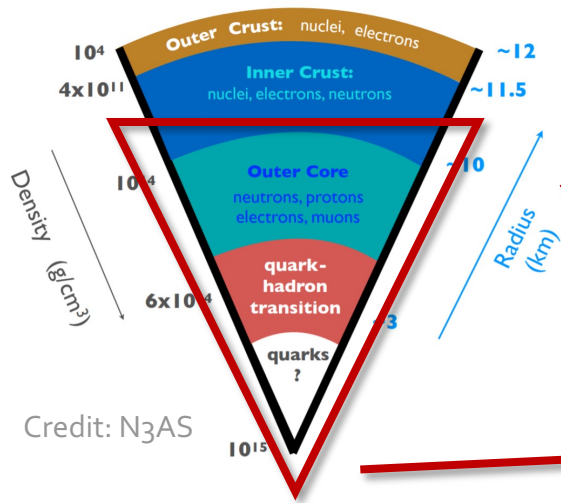
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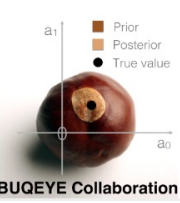
Bayes your way to victory!

Halloween 2024





# “Low” densities: EOS from chiral EFT



QCD non-perturbative at low energies, build *effective description* using nucleons, pions as degrees of freedom

$$Q = \max \left( \frac{p}{\Lambda_b}, \frac{m_\pi}{\Lambda_b} \right)$$

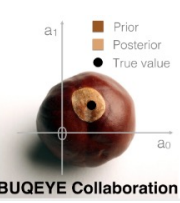
Quantifiable truncation error, obeys all symmetries of QCD

	NN forces	3N forces
LO (Q <sup>0</sup> )	(1990) <span style="float: right;">2</span>	—
NLO (Q <sup>2</sup> )	(1992) <span style="float: right;">7</span>	— (1992 94)
N <sup>2</sup> LO (Q <sup>3</sup> )	(1992) <span style="float: right;">0</span>	(1994) <span style="float: right;">2</span>
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$$\mathcal{E}_2(k_F) = \text{diagram} = -\frac{1}{4} \sum_{ijab} V^{ij,ab} V^{ab,ij} f_{ij} \bar{f}_{ab} \frac{1}{D_{ab,ij}}$$

$$\mathcal{E}_{3,pp}(k_F) = \text{diagram}$$

Use **MBPT\*** for energy per particle (E/A) calculations

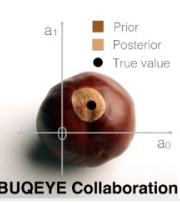
\*A choice, not a necessity--- BMM framework is model-independent

$$\langle \mathbf{2}' \mathbf{3}' | V_{NN}^{\text{med}} | \mathbf{23} \rangle = \text{diagram}$$

C. Drischler, J. Holt, C. Wellenhofer (2021)



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C. Drischler, S. Bogner (2021)

Using NN potential from **Entem, Machleidt, Nosyk (2017)** + **3N forces**  
 Fit **low-energy constants (LECs)** to experimental data, empirical saturation point of SNM  
 Employ momentum **cutoff** of 500 MeV

$$\mathcal{E}_2(k_F) = \left[ \text{diagram} \right] = -\frac{1}{4} \sum_{ijab} V^{ij,ab} V^{ab,ij} f_{ij} \bar{f}_{ab} \frac{1}{D_{ab,ij}}$$

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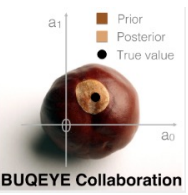
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# Uncertainty quantification for the EOS

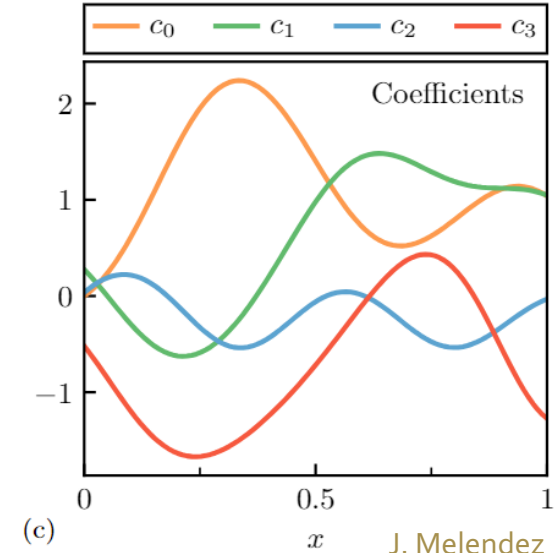
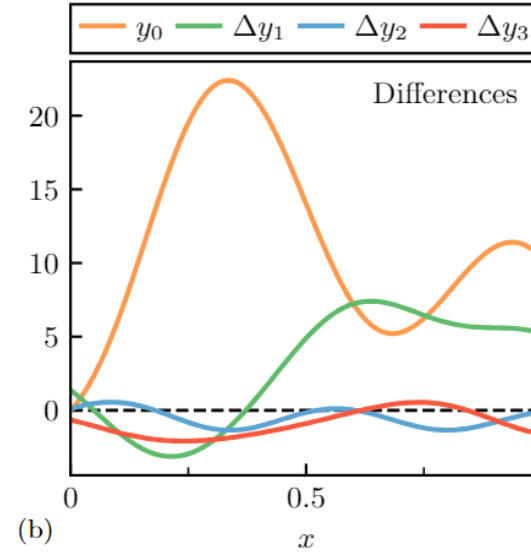
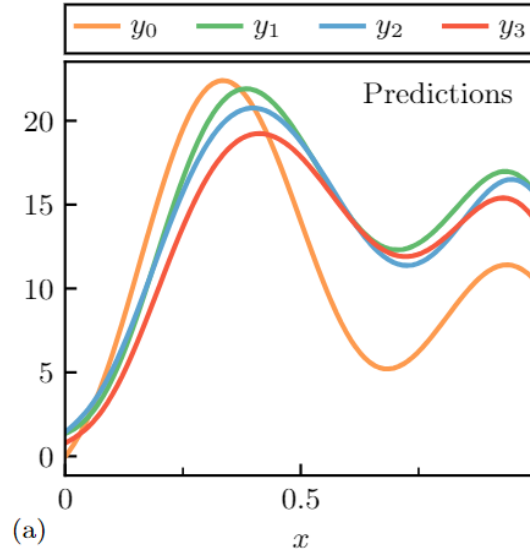


Employ the BUQEYE truncation error model (**gsum**)



$$y_k(x) = y_{\text{ref}}(x) \sum_{n=0}^k c_n(x) Q^n(x)$$

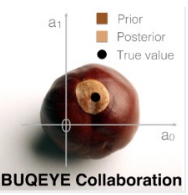
Known orders of expansion



J. Melendez et al. (2019)



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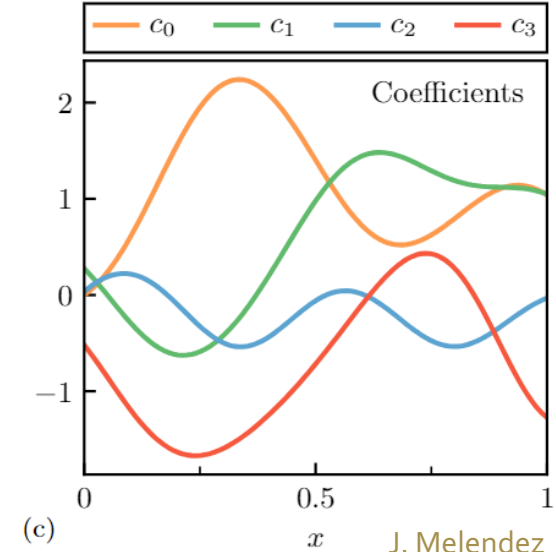
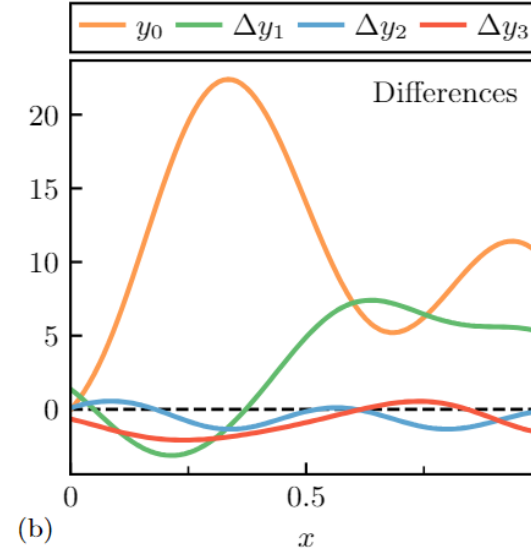
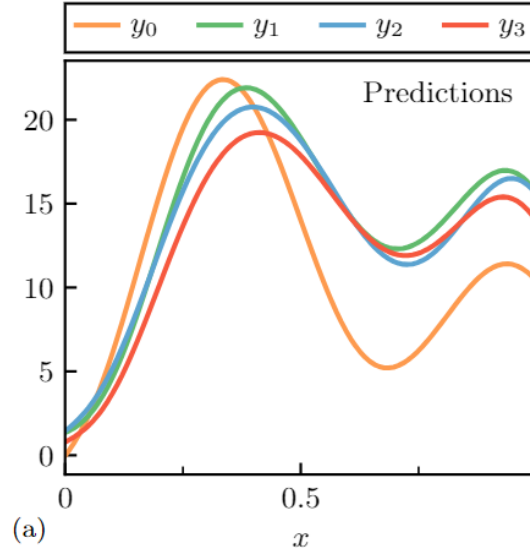


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Estimations for unknown orders

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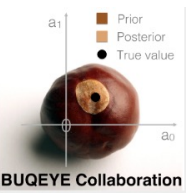
Dimensional analysis

Natural i.i.d. curves

Expansion parameter (suppresses higher orders)



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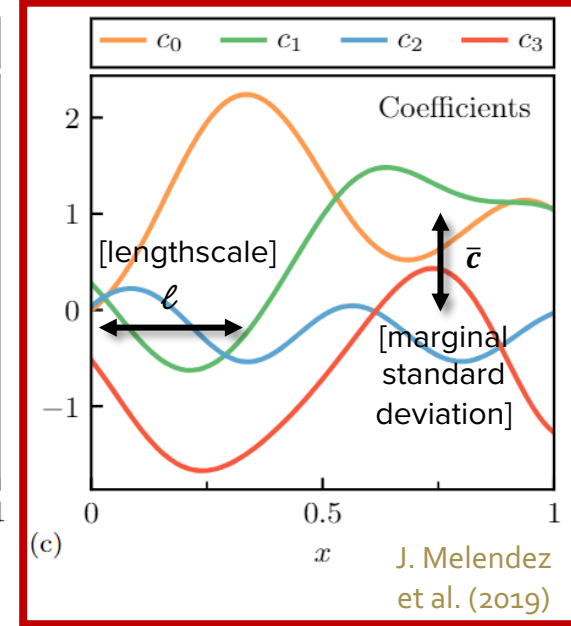
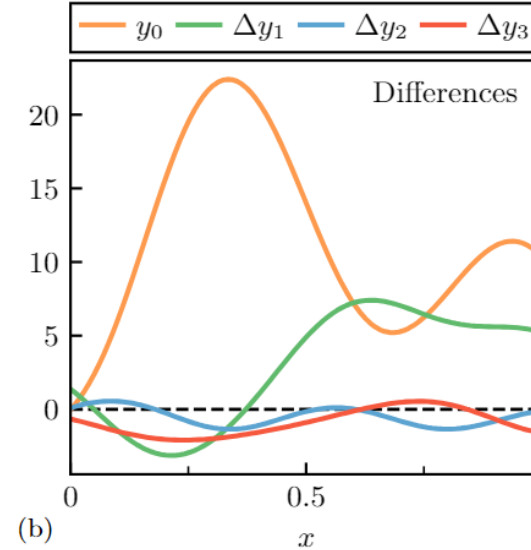
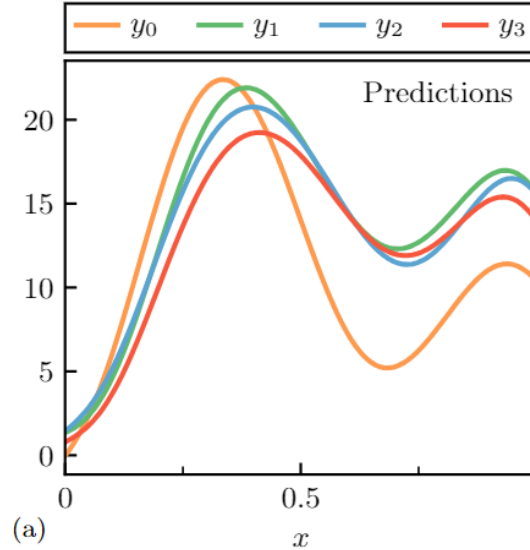
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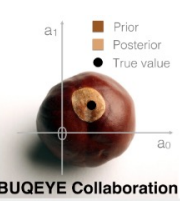
**Gaussian Processes:** learn coefficient distributions

Incorporates correlations in  $y(x)$ , continuous model, priors constrain parameters of GP

Easily test with *diagnostics* (Mahalanobis distance, pivoted Cholesky decomposition,...)



# “Low” densities: EOS from chiral EFT



Obtain pressure as a function of number density,  $P(n)$ , for model mixing calculations

$$P(n) = n^2 \frac{d}{dn} \frac{E}{A}(n)$$

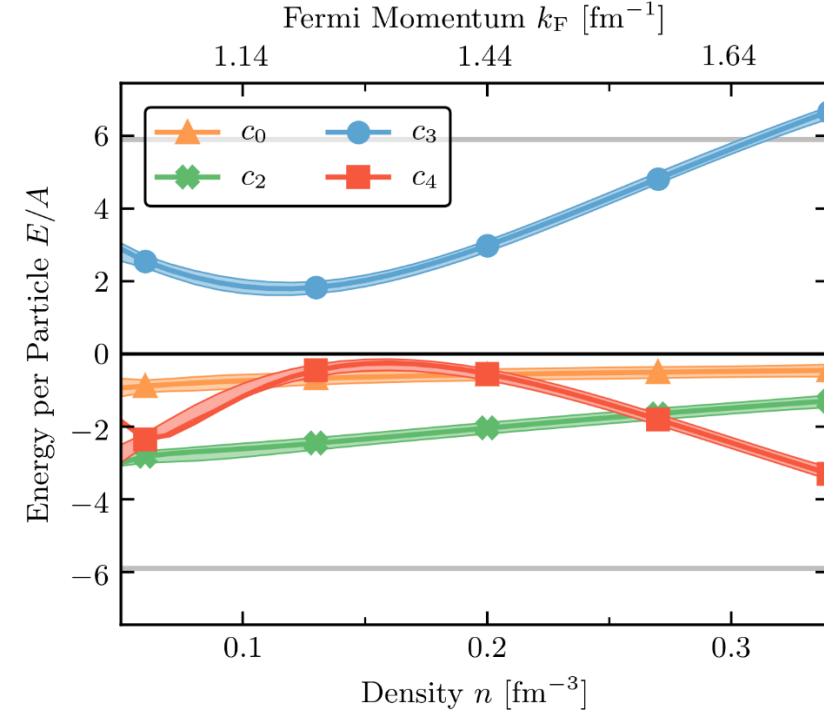
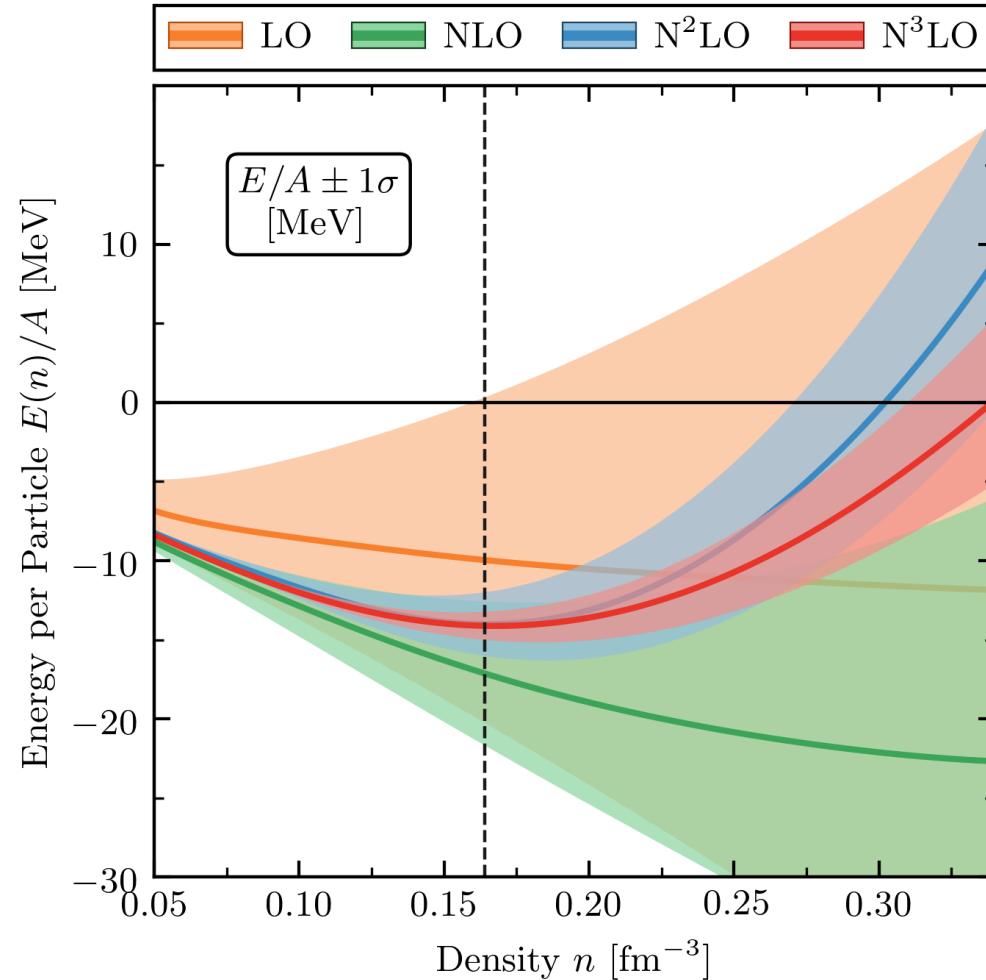


Coefficient extraction for truncation error estimation done via **gsum**

$$Q(k_F) = \frac{k_F}{\Lambda_b} \rightarrow \approx 600 \text{ MeV}$$

$$y_{\text{ref}}(k_F) = 16 \text{ MeV} \times \left( \frac{k_F}{k_{F,0}} \right)^2$$

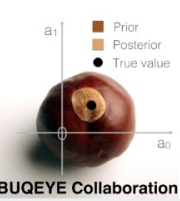
*Symmetric nuclear matter*



Truncation error scheme yields natural-sized curves as expected



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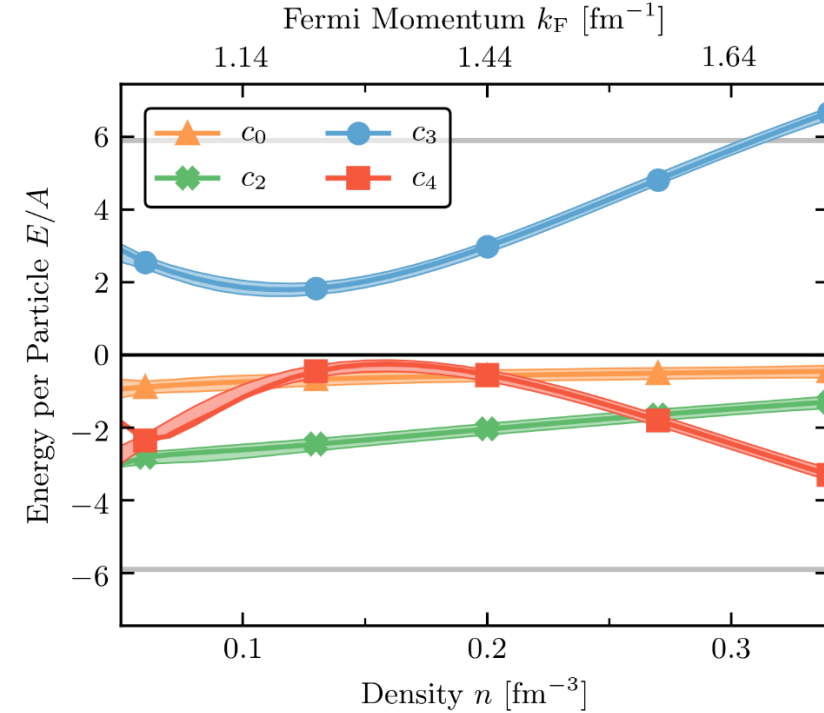
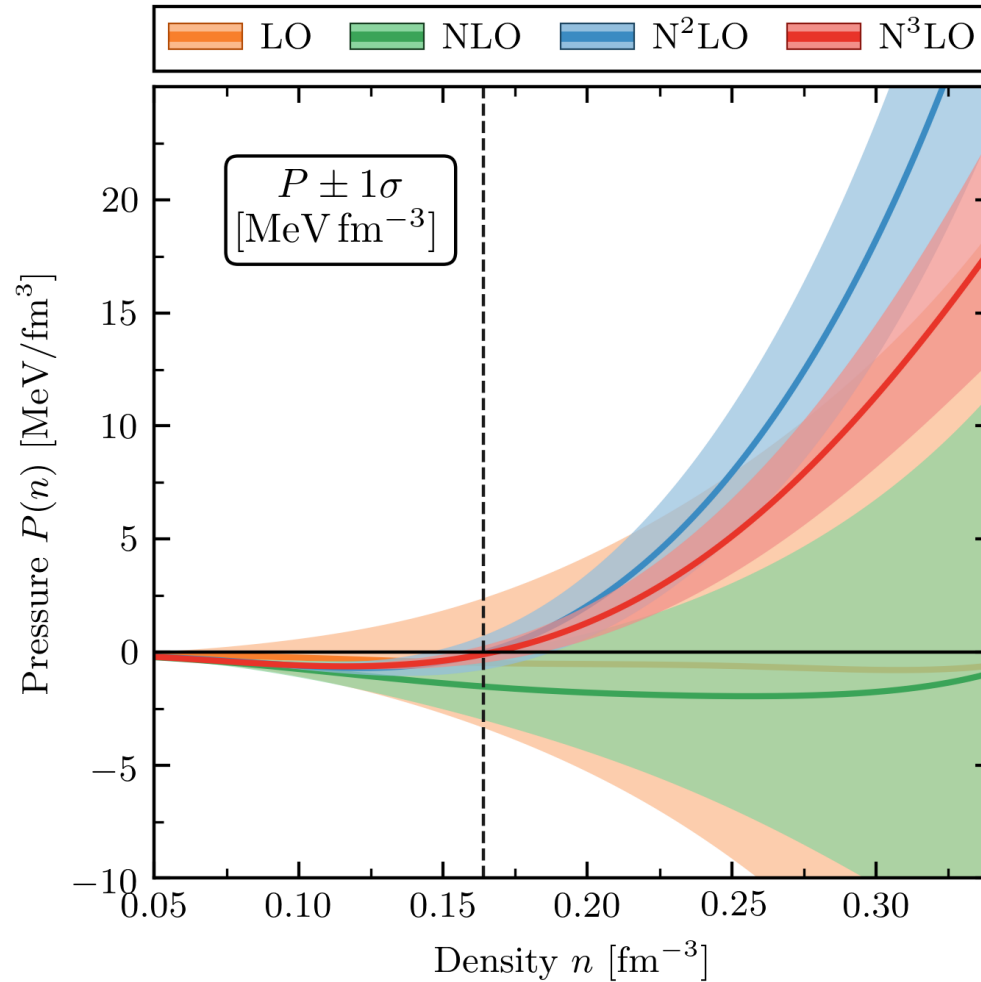


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# High densities: pQCD EOS

## QCD perturbative at high energies

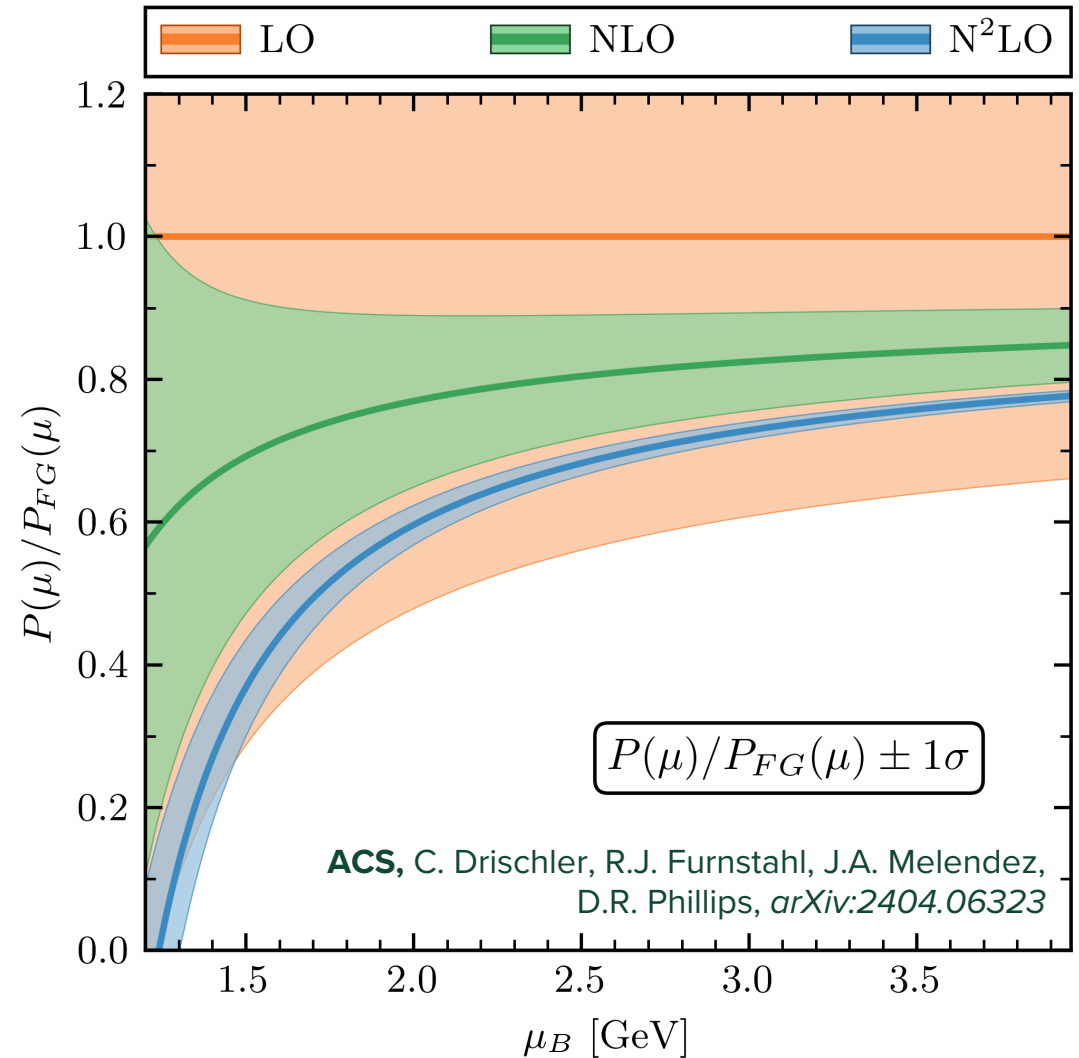
Expansion in the strong coupling constant  $\alpha_s$

$$\begin{aligned} \frac{P(\mu)}{P_{FG}(\mu)} \simeq & 1 + a_{1,1} \left( \frac{\alpha_s(\mu)}{\pi} \right) \\ & + N_f \left( \frac{\alpha_s(\mu)}{\pi} \right)^2 \left[ a_{2,1} \ln \left( \frac{N_f \alpha_s(\mu)}{\pi} \right) \right. \\ & \left. + a_{2,2} \ln \frac{\bar{\Lambda}}{2\mu} + a_{2,3} \right] + \mathcal{O}(\alpha_s^3), \end{aligned}$$

Two-loop running:

$$\alpha_s(\bar{\Lambda}) = \frac{4\pi}{\beta_0 L} \left[ 1 - \frac{2\beta_1}{\beta_0^2} \frac{\ln L}{L} \right] \begin{cases} L = \ln(\bar{\Lambda}^2 / \Lambda_{MS}^2), \\ \bar{\Lambda} = 2X\mu \end{cases}$$

Degrees of freedom: quarks and gluons  
Massless u, d quarks with equal  $\mu$



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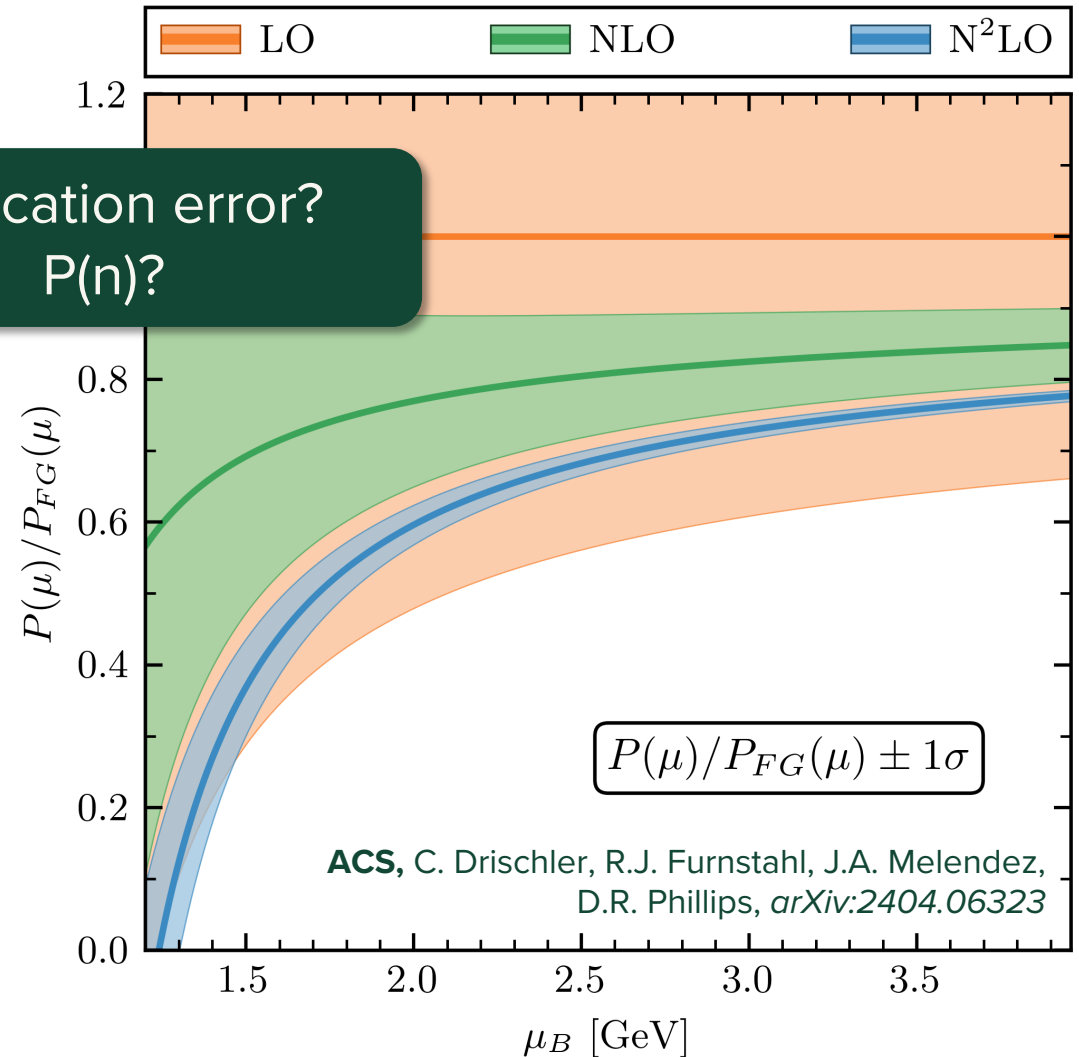
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Truncation error?  
P(n)?



ACS, C. Drischler, R.J. Furnstahl, J.A. Melendez, D.R. Phillips, arXiv:2404.06323

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Invoke the Kohn-Luttinger-Ward inversion theorem:  $P(\mu) \rightarrow P(n)$

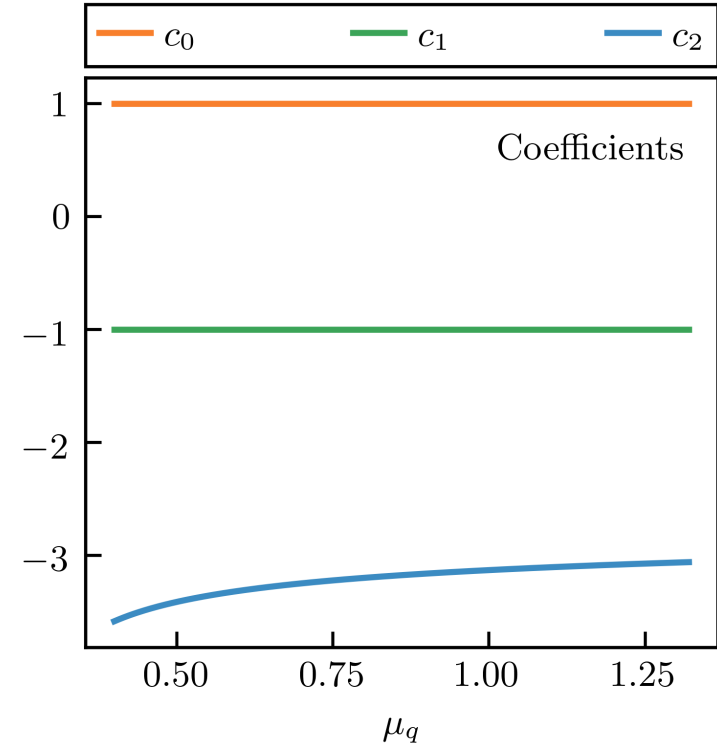
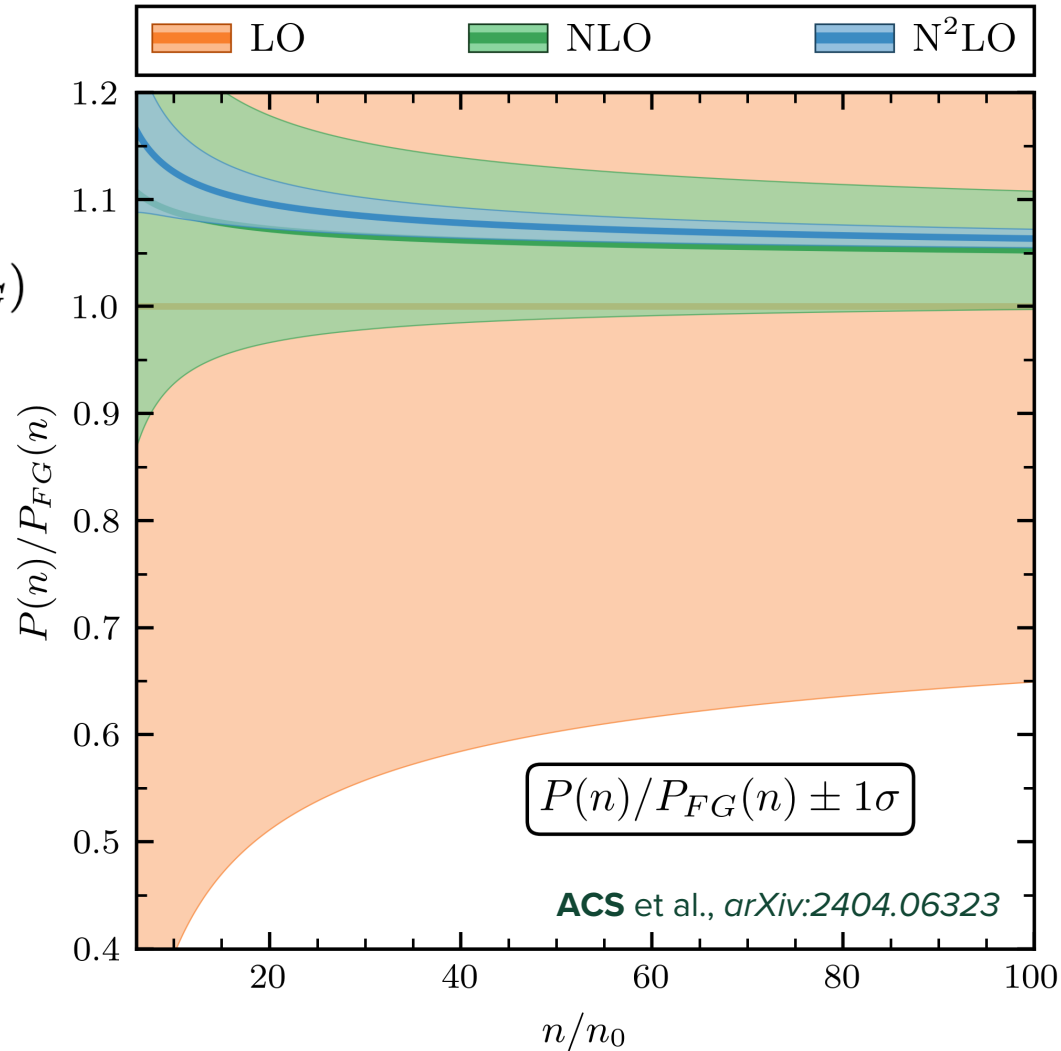
$$P(n) = P_{FG}(n) \left[ c_0 + c_1 Q(\bar{\Lambda}_{FG}) + c_2(n) Q^2(\bar{\Lambda}_{FG}) + \dots \right]$$

Apply **gsum** to **P(n)**

$$Q = \frac{N_f}{\pi} \alpha_s(\bar{\Lambda})$$



$$y_{\text{ref}} = P_{FG}(n)$$



**Truncation error** often assessed by varying renormalization scale to obtain band---not statistically rigorous

\* See newer works by Gorda et al. (2022, 2023), MiHO model for higher orders



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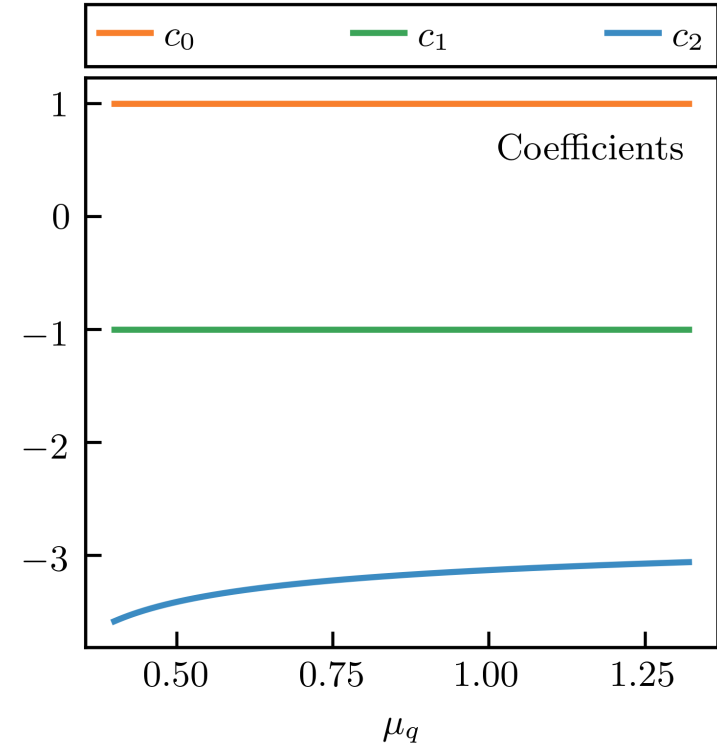
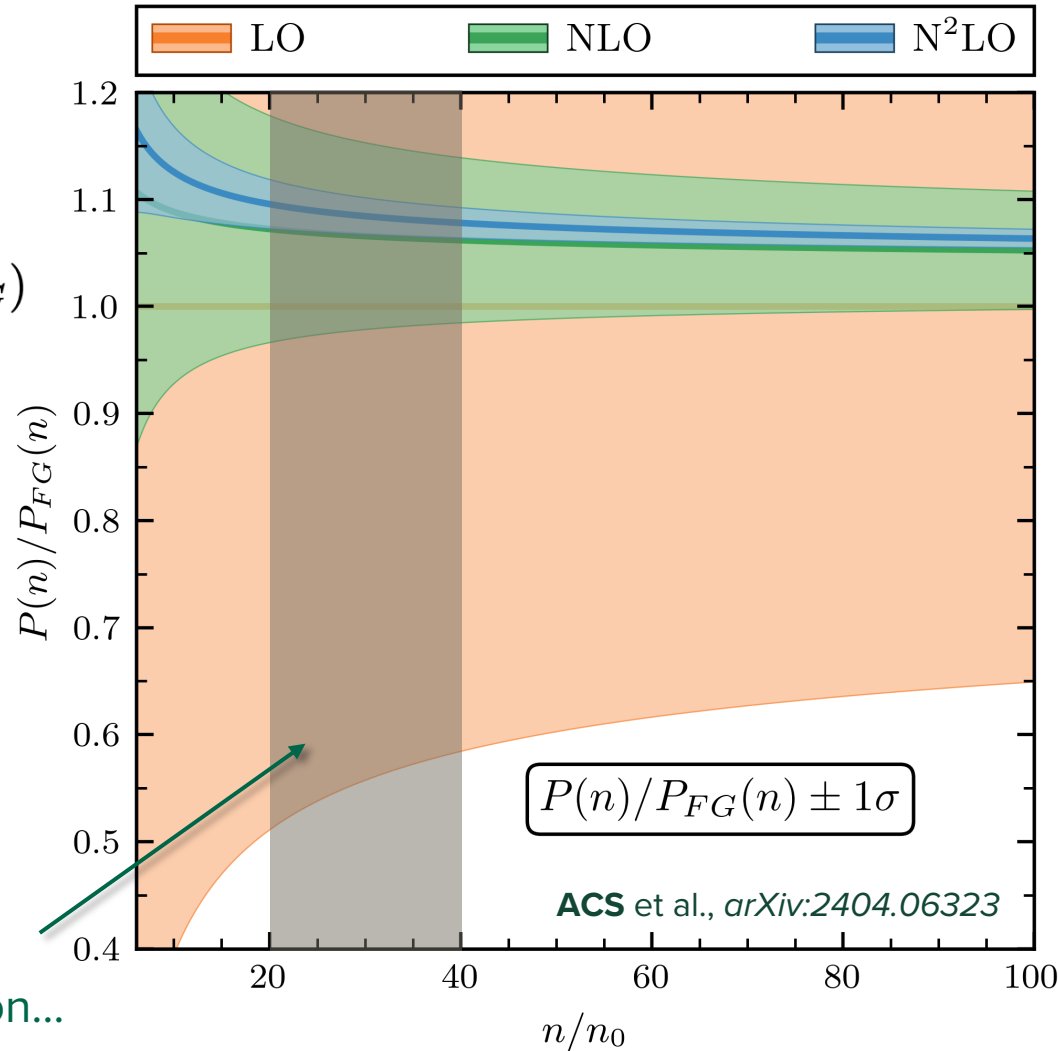
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Nonperturbative effects under (20-40)  $n_0$ : pairing, hadronization...



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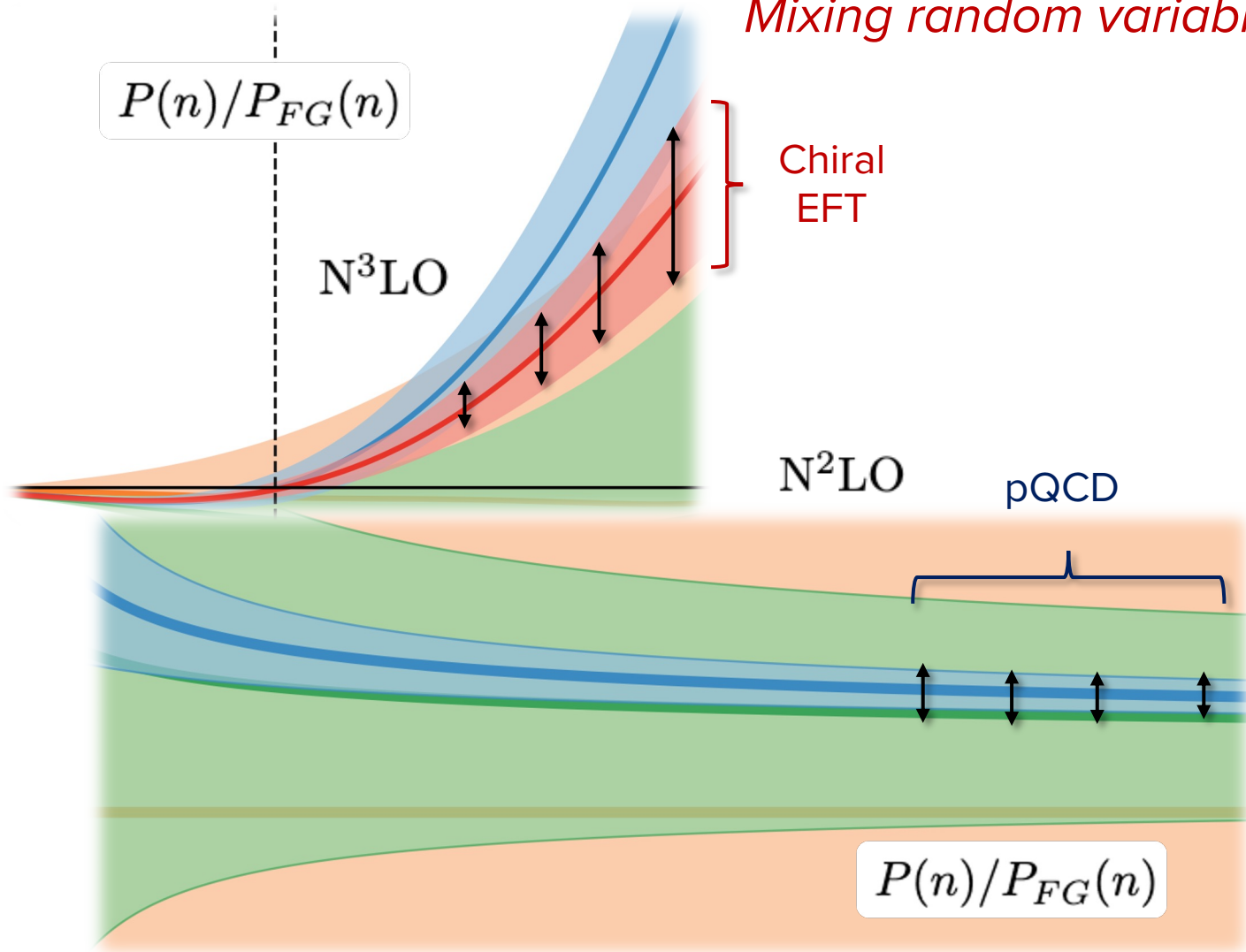


# GPs: a correlated approach to BMM

Mixing random variables "curvewise"

Underlying theory (QCD)

$$Y_i = F + \delta Y_i, \quad i \in [0, M]$$

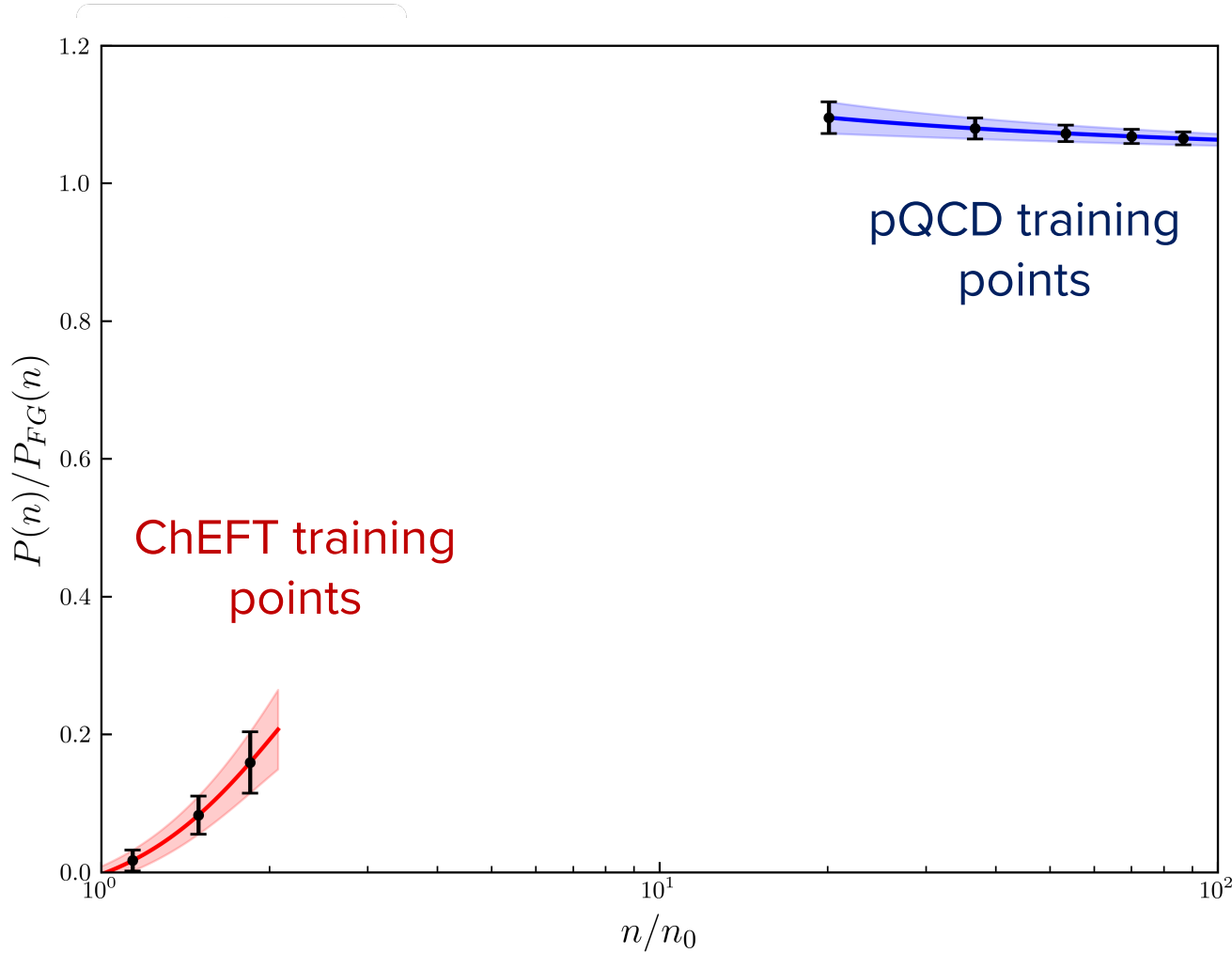


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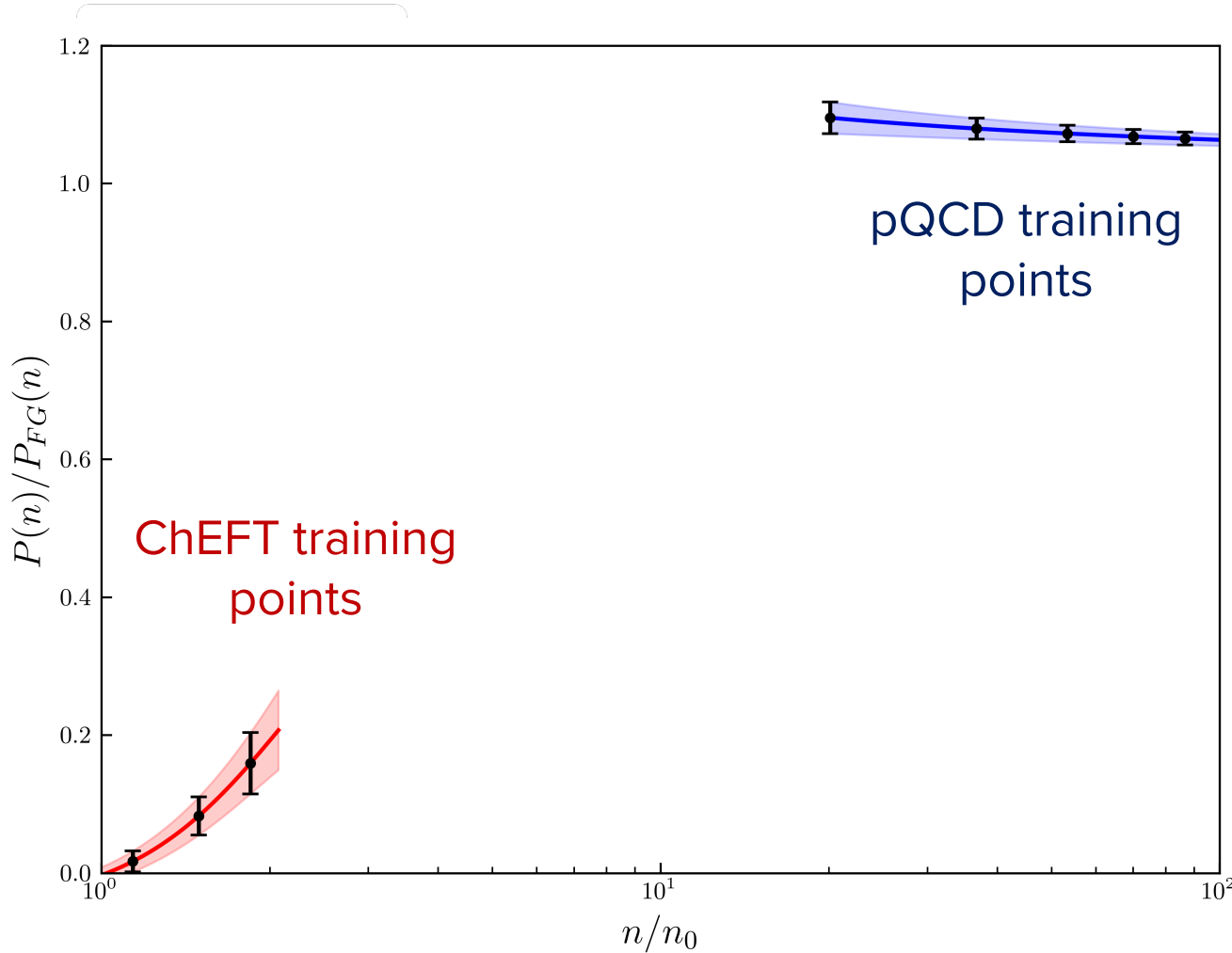
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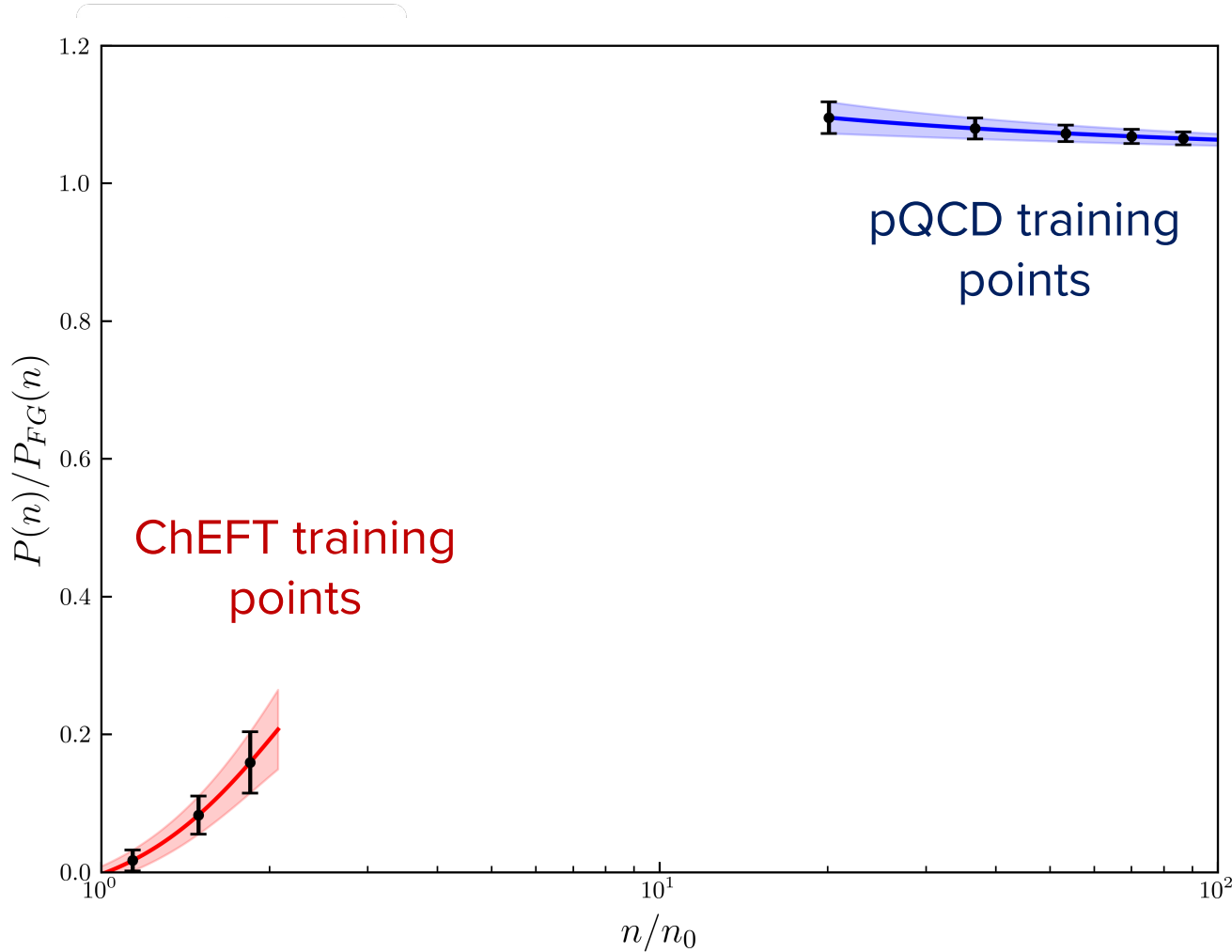
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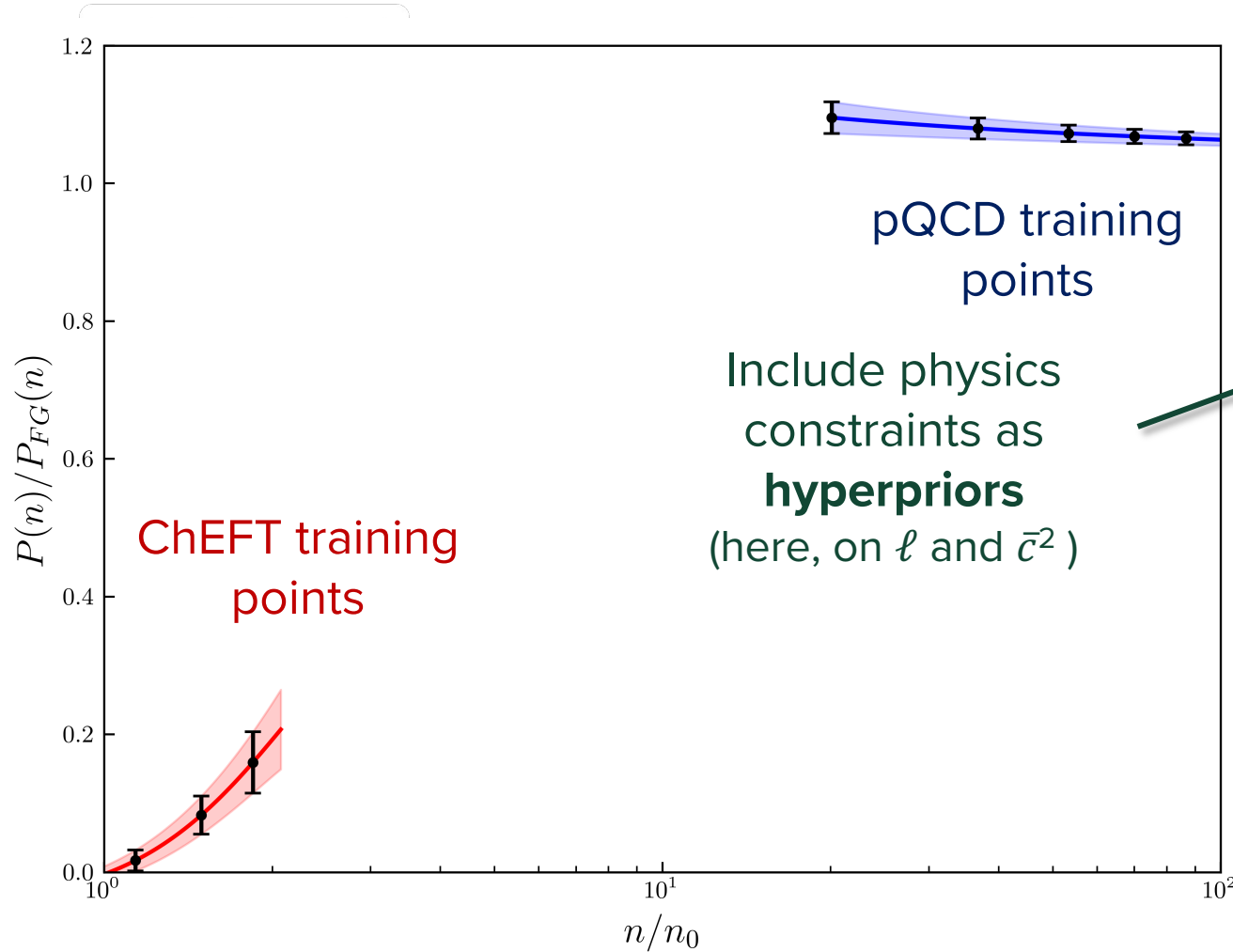
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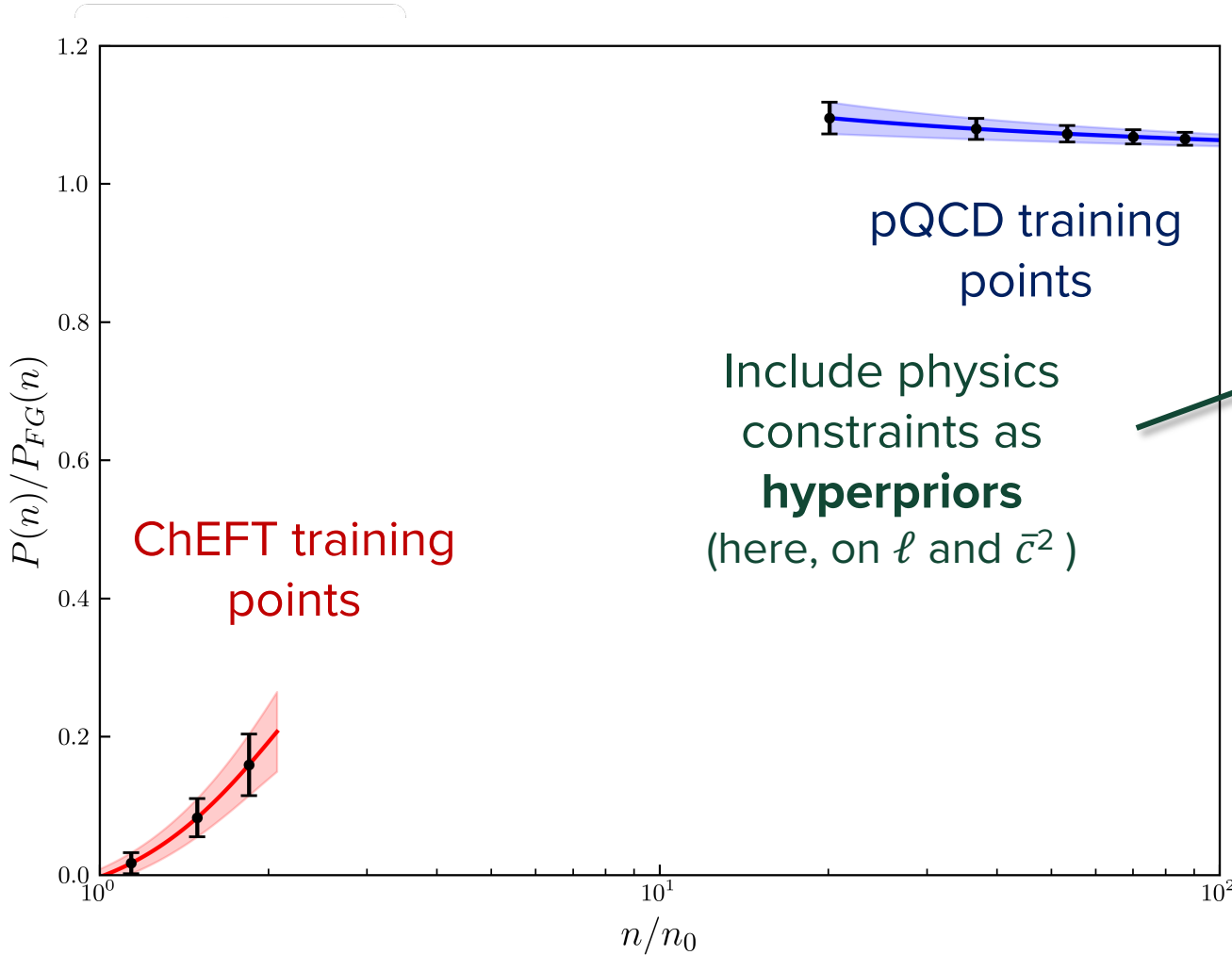
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$$\sim \text{GP}[0, \kappa_y^{(i)}(x, x')] \quad [\text{Full covariance matrix}]$$

Therefore, assuming a **Gaussian form** and using Bayes' theorem, again determine a **common mean**

$$F | \vec{y}, K_y, K_f \sim \mathcal{N}[\mu, \Sigma]$$

$$\vec{\mu} \equiv \Sigma B_t^T K_y^{-1} \vec{y}$$

$$\Sigma \equiv (K_f^{-1} + B_t^T K_y^{-1} B_t)^{-1}$$



# GPs: a correlated approach to BMM

Mixing random variables “curvewise”

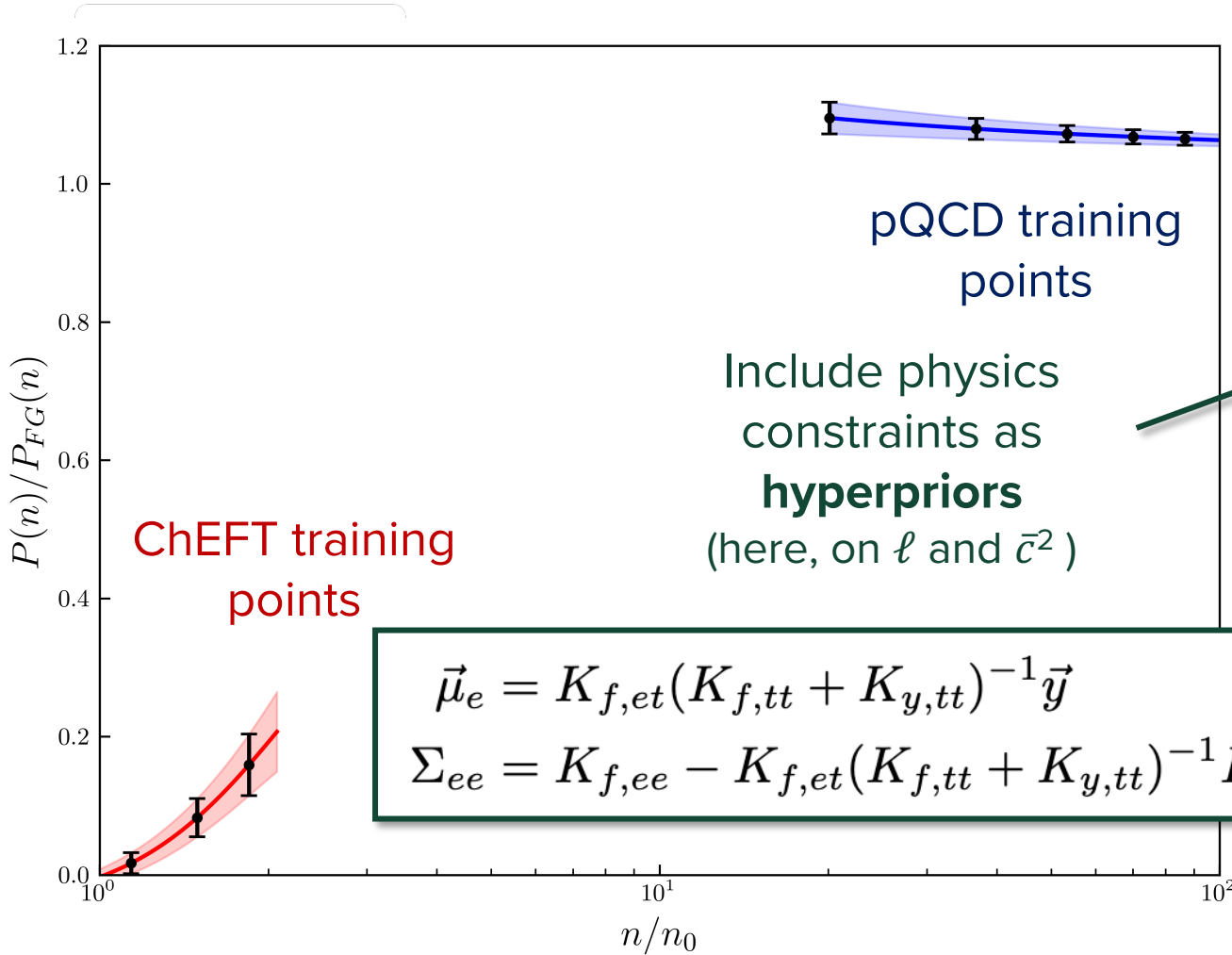
Underlying theory (QCD)

$$Y_i = F + \delta Y_i, \quad i \in [0, M]$$

~ GP[0,  $\kappa_f(x, x')$ ]  
[Kernel choice]

~ GP[0,  $\kappa_y^{(i)}(x, x')$ ]  
[Full covariance matrix]

Therefore, assuming a **Gaussian form** and using Bayes’ theorem, again determine a **common mean**



$$\vec{\mu}_e = K_{f,et}(K_{f,tt} + K_{y,tt})^{-1}\vec{y}$$

$$\Sigma_{ee} = K_{f,ee} - K_{f,et}(K_{f,tt} + K_{y,tt})^{-1}K_{f,te}$$

$$F | \vec{y}, K_y, K_f \sim \mathcal{N}[\mu, \Sigma]$$

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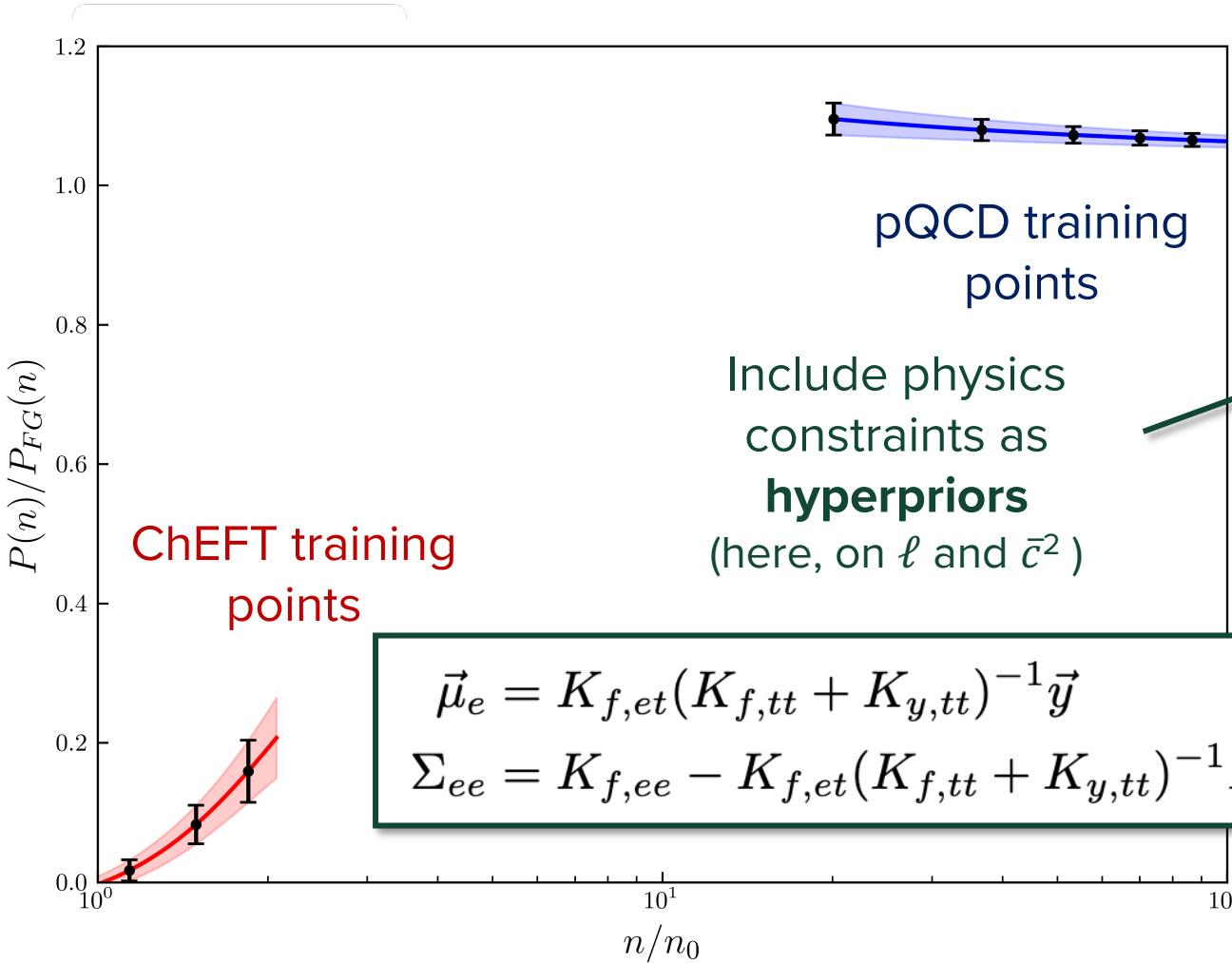
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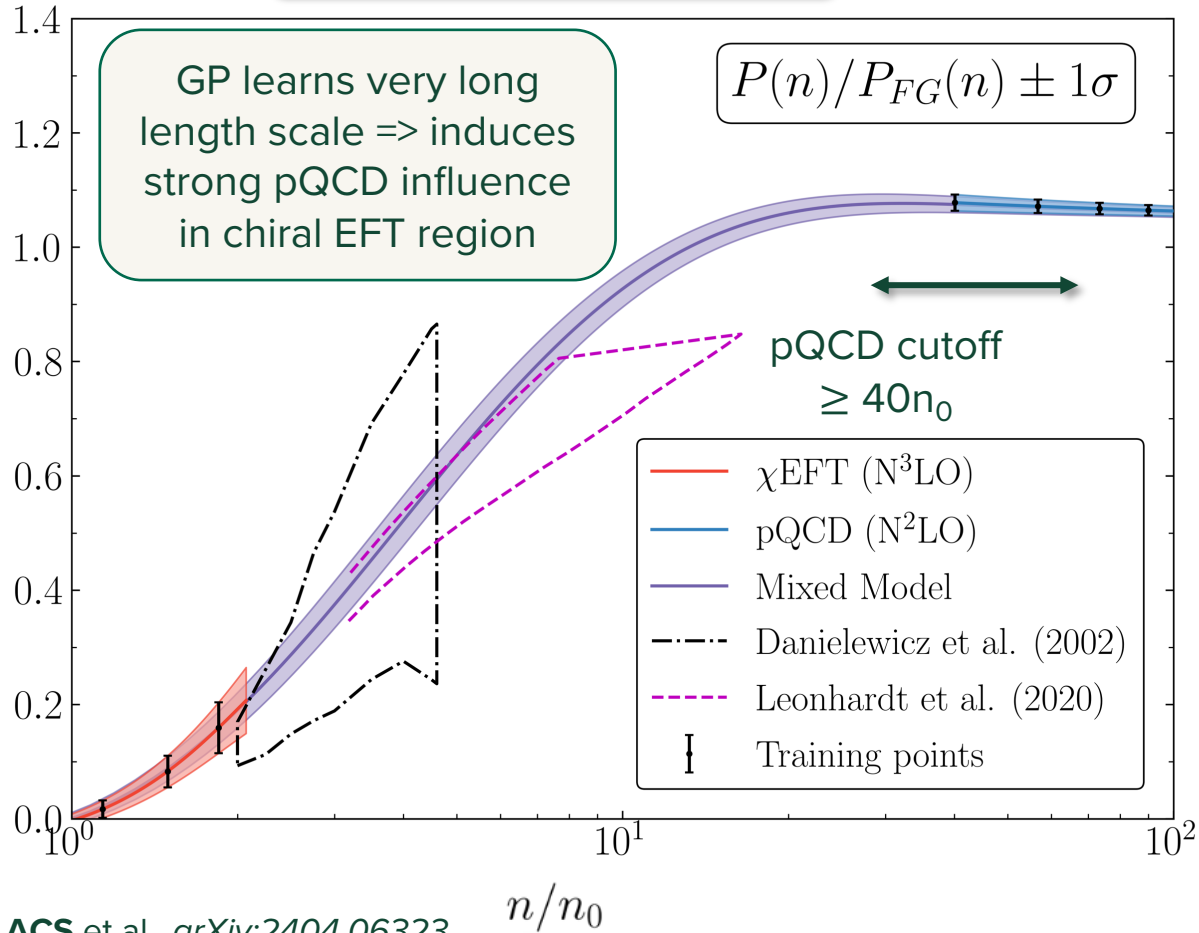
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**GP equations are the mixed model result!**



# Results: pressure of SNM

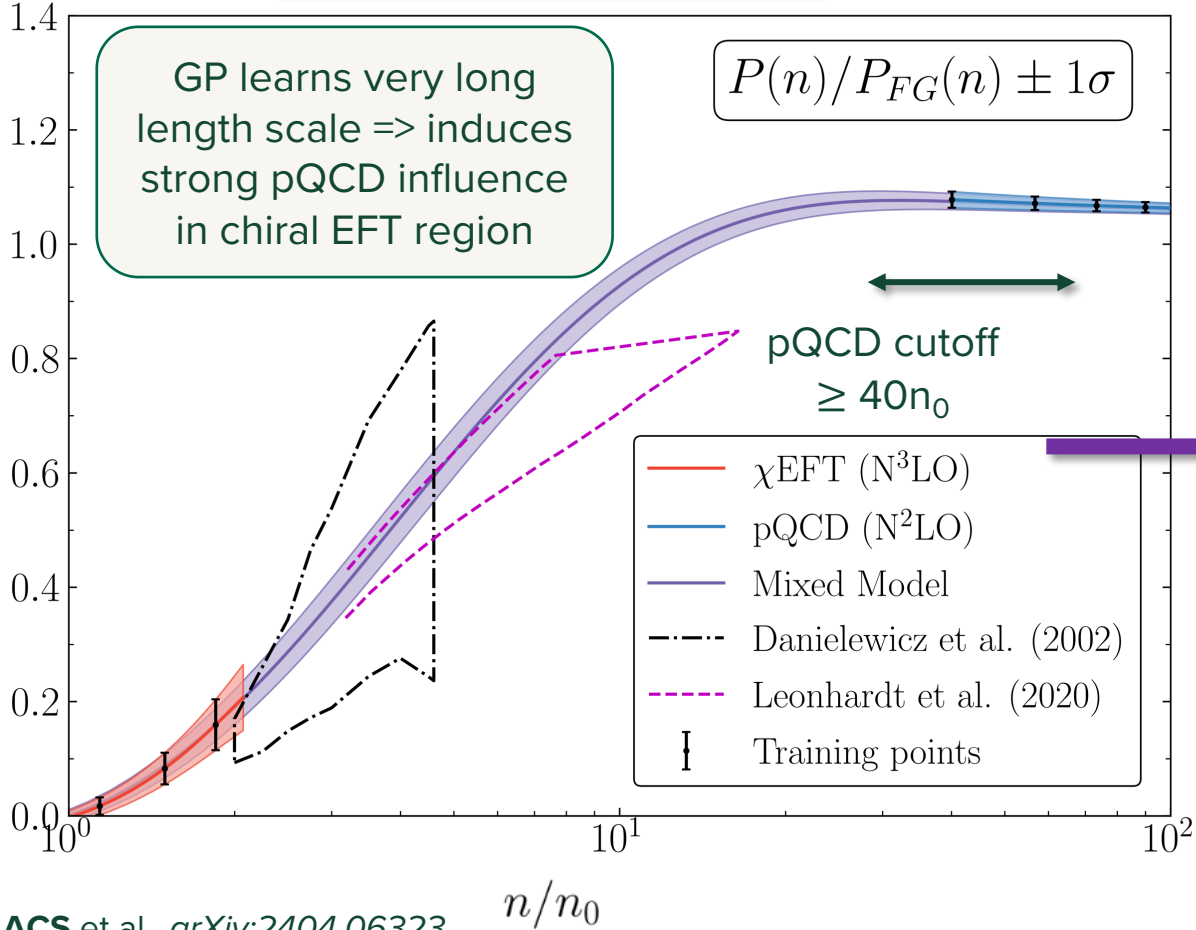
Unconstrained prior



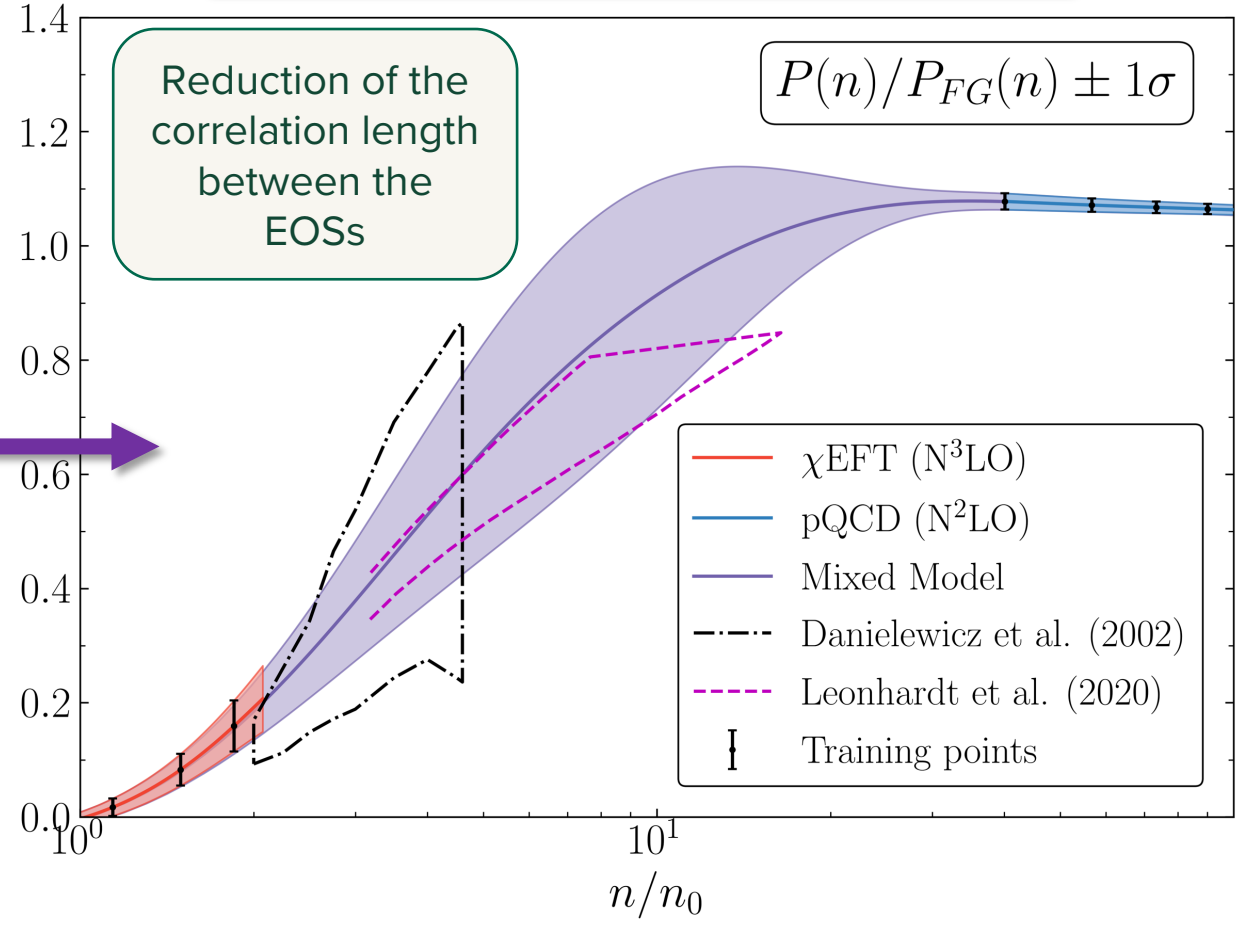
ACS et al., arXiv:2404.06323

# Results: pressure of SNM

Unconstrained prior



Constrained prior on lengthscale



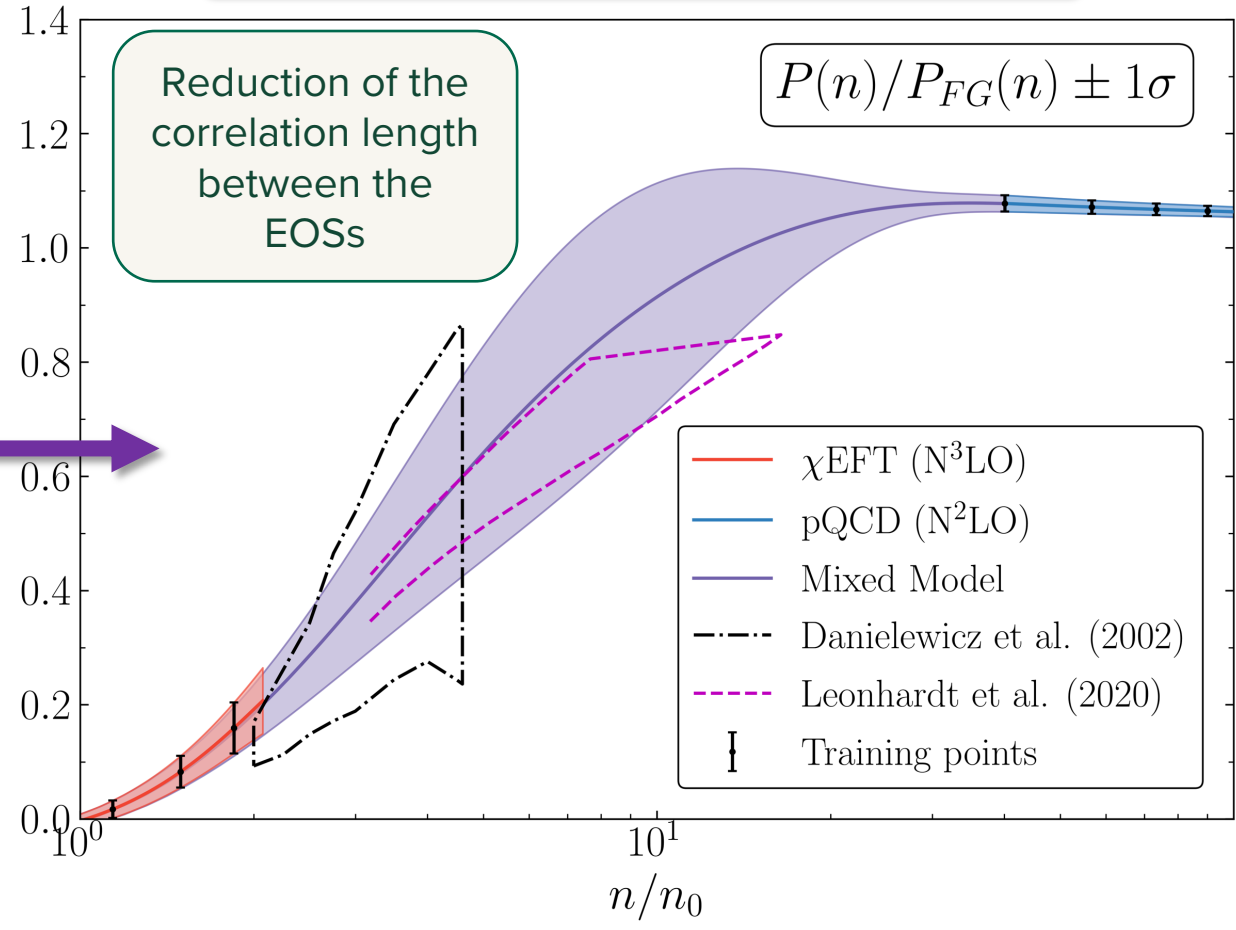
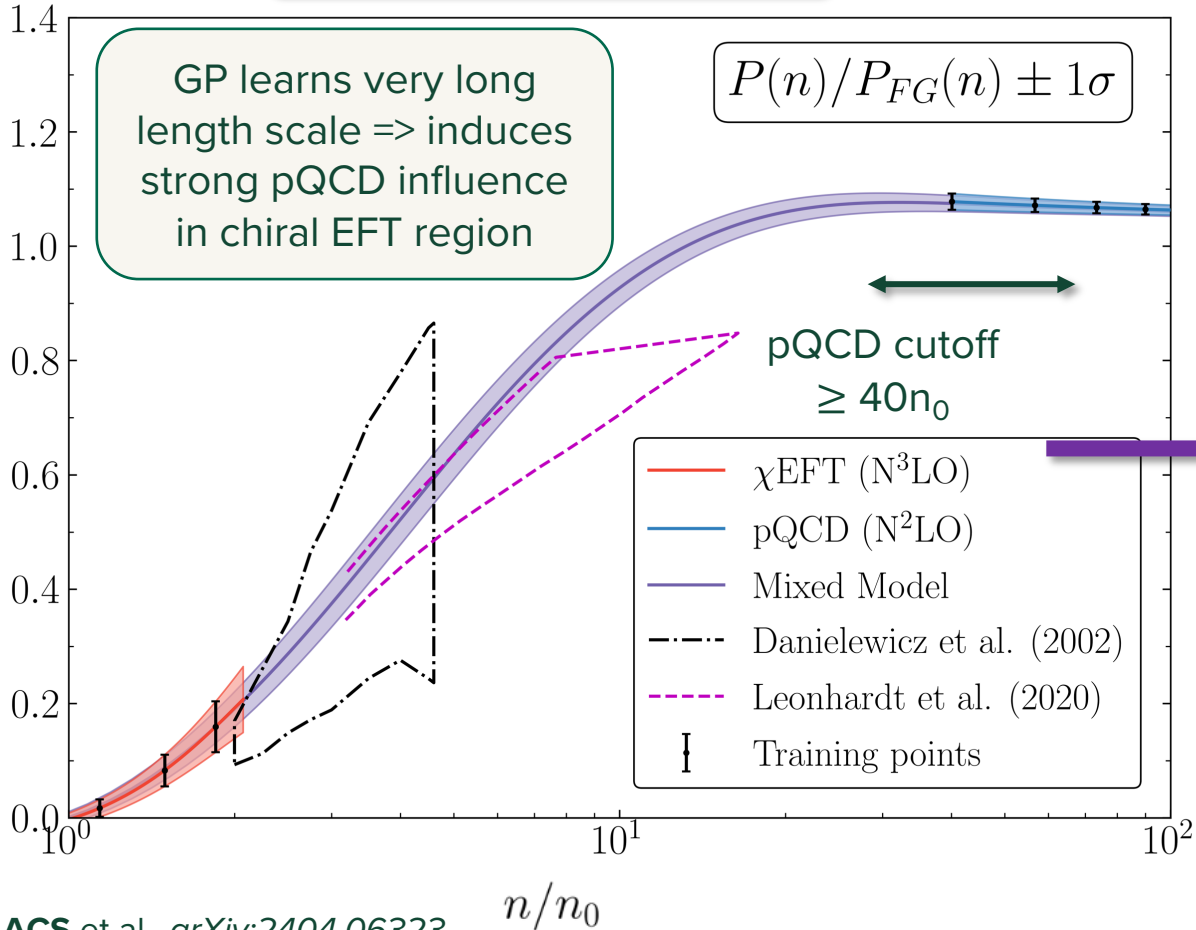
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# Results: pressure of SNM

Unconstrained prior

Constrained prior on lengthscale



ACS et al., arXiv:2404.06323

Physics-informed priors on the mixed model GP have a very large influence on the uncertainties obtained



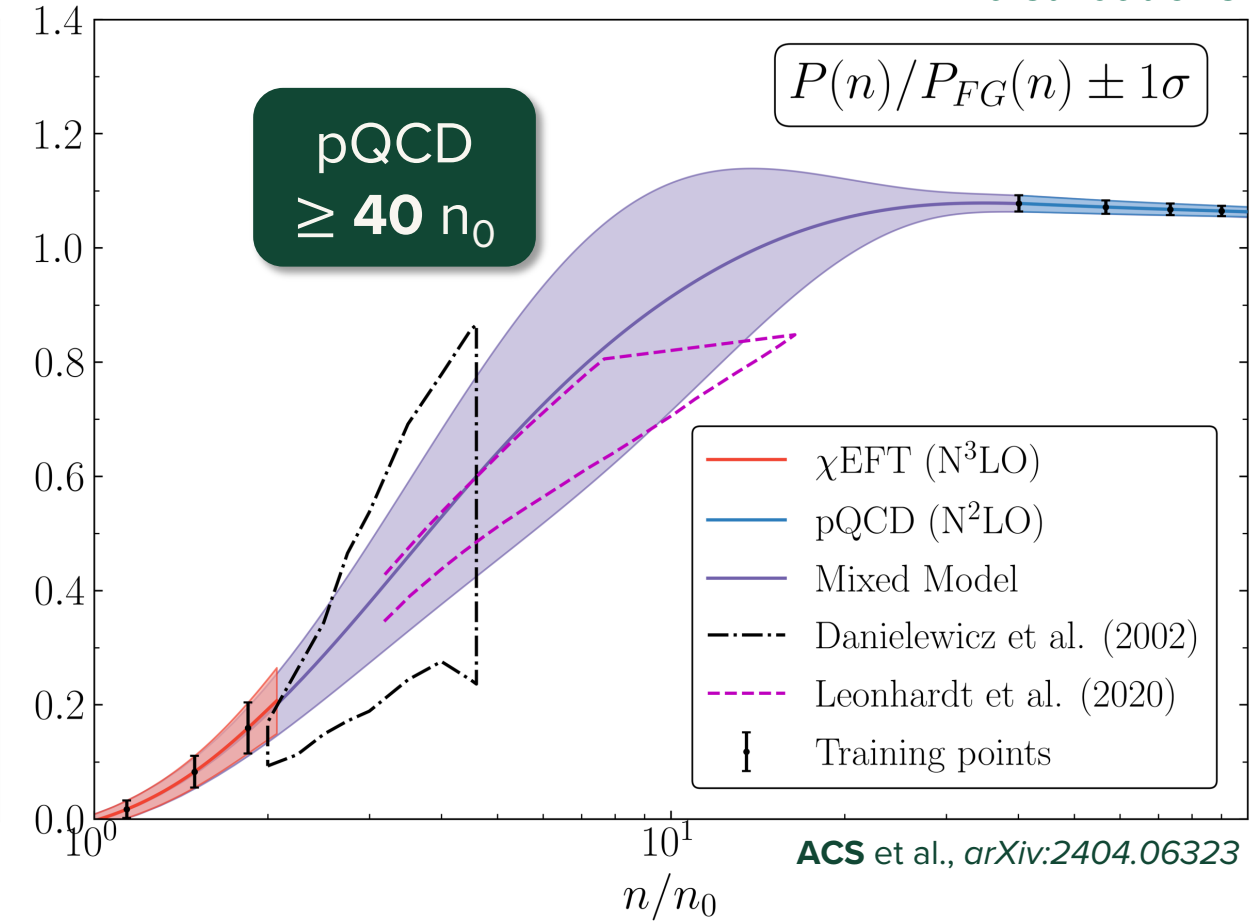
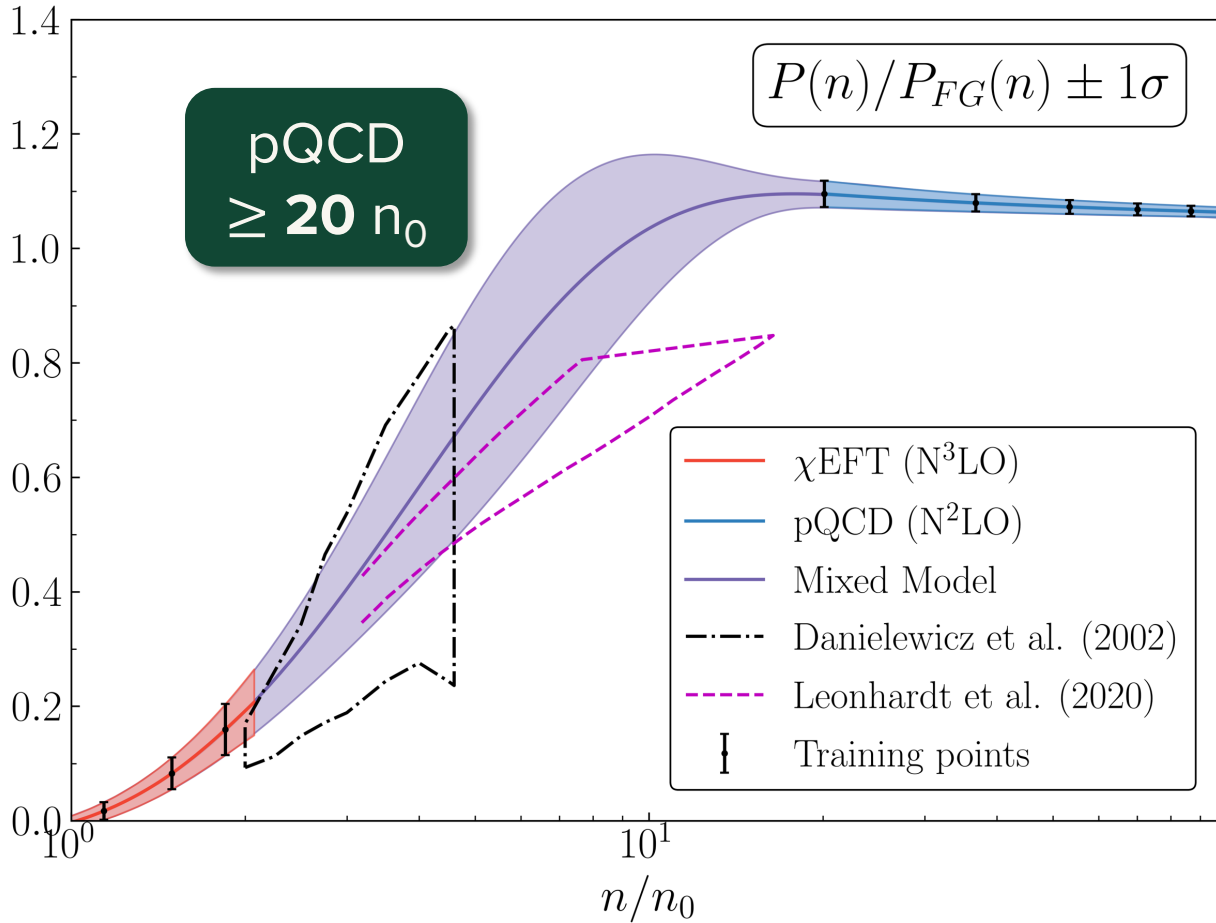


# Results: pressure

Unknown cutoff of pQCD validity = testing with **two cases**: pQCD cutoffs for  $\geq 20 n_0$  &  $\geq 40 n_0$

**Hyperpriors**  
truncated normal distributions

$$p(\vec{\theta}) = \prod_i \mathcal{U}(\theta_i \in [a_i, b_i]) \mathcal{N}(\theta_i, \mu_i, \sigma_i^2)$$



ACS et al., arXiv:2404.06323

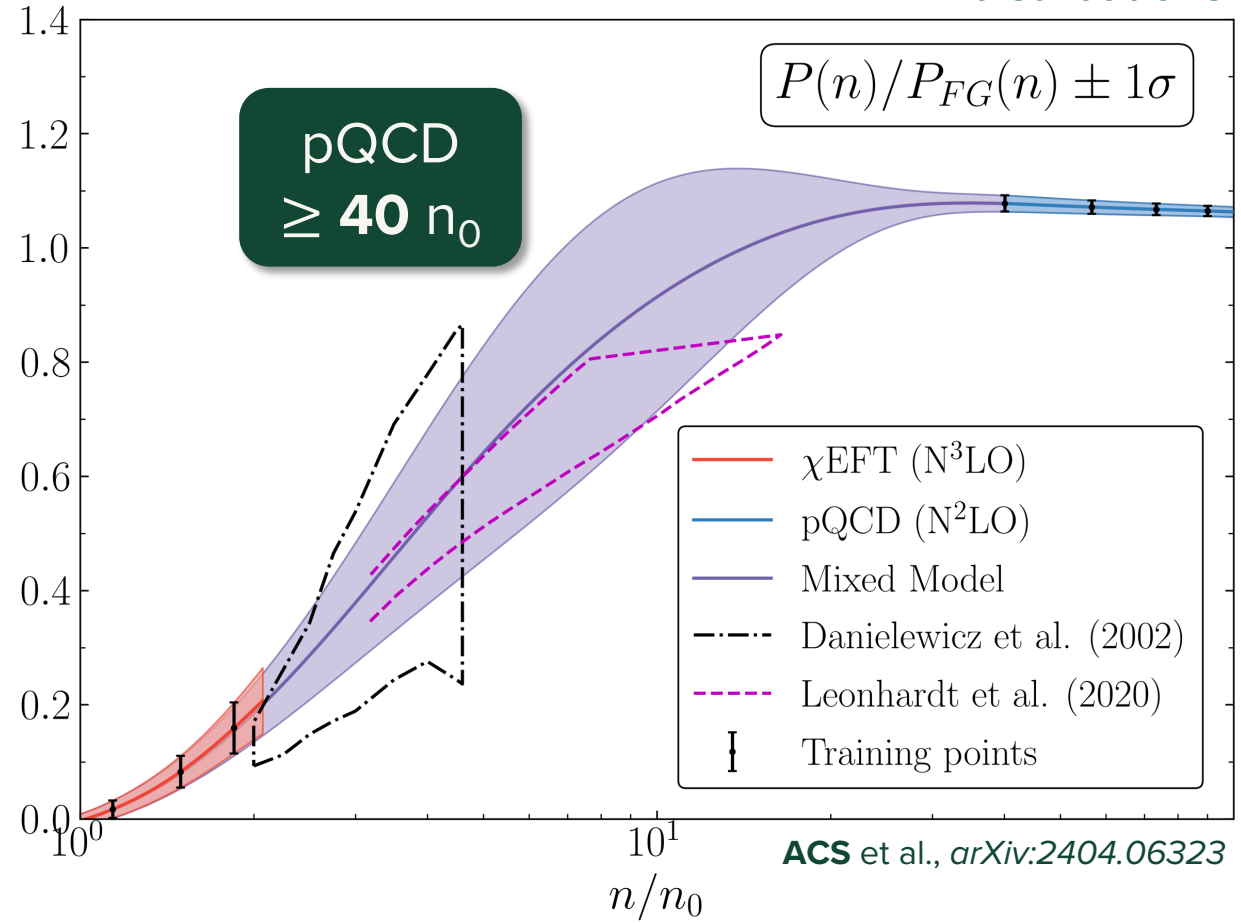
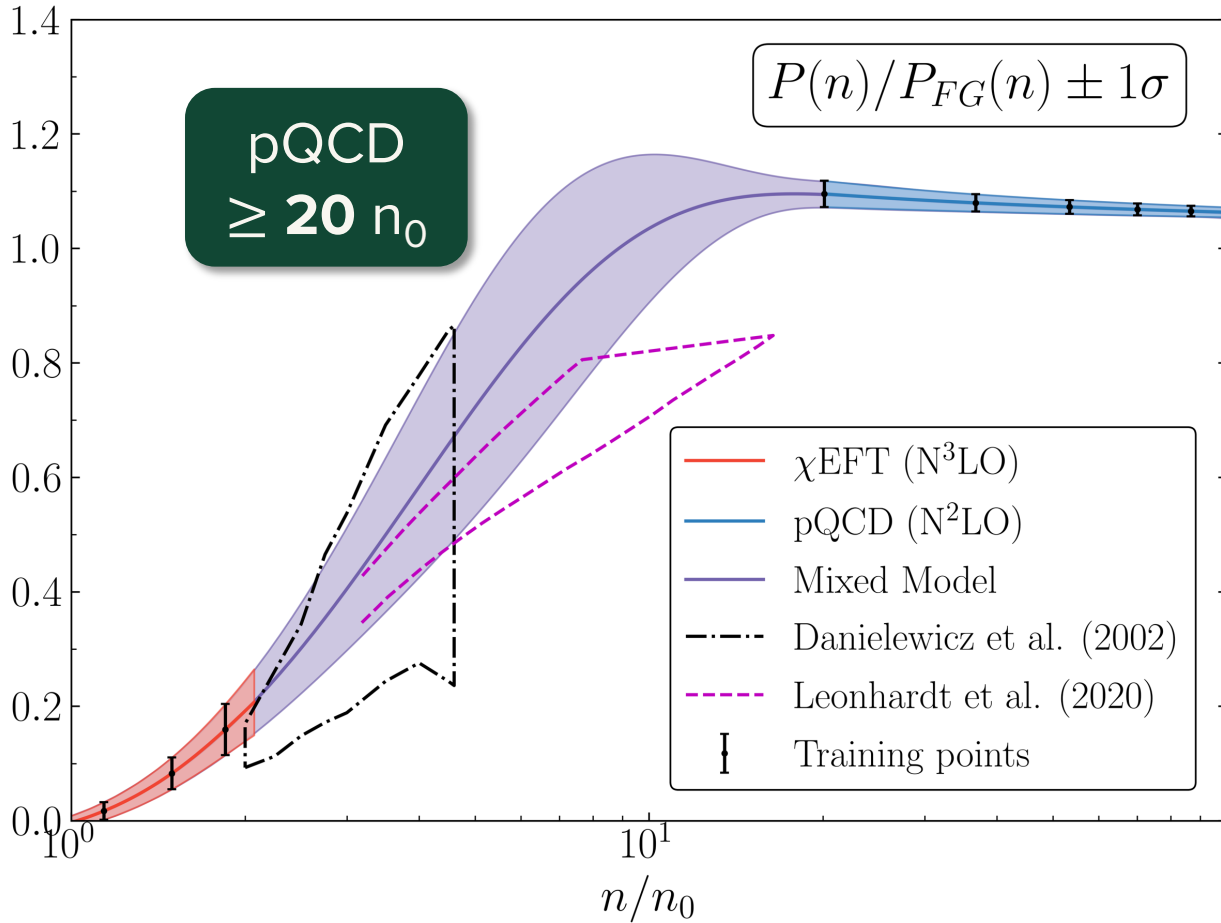


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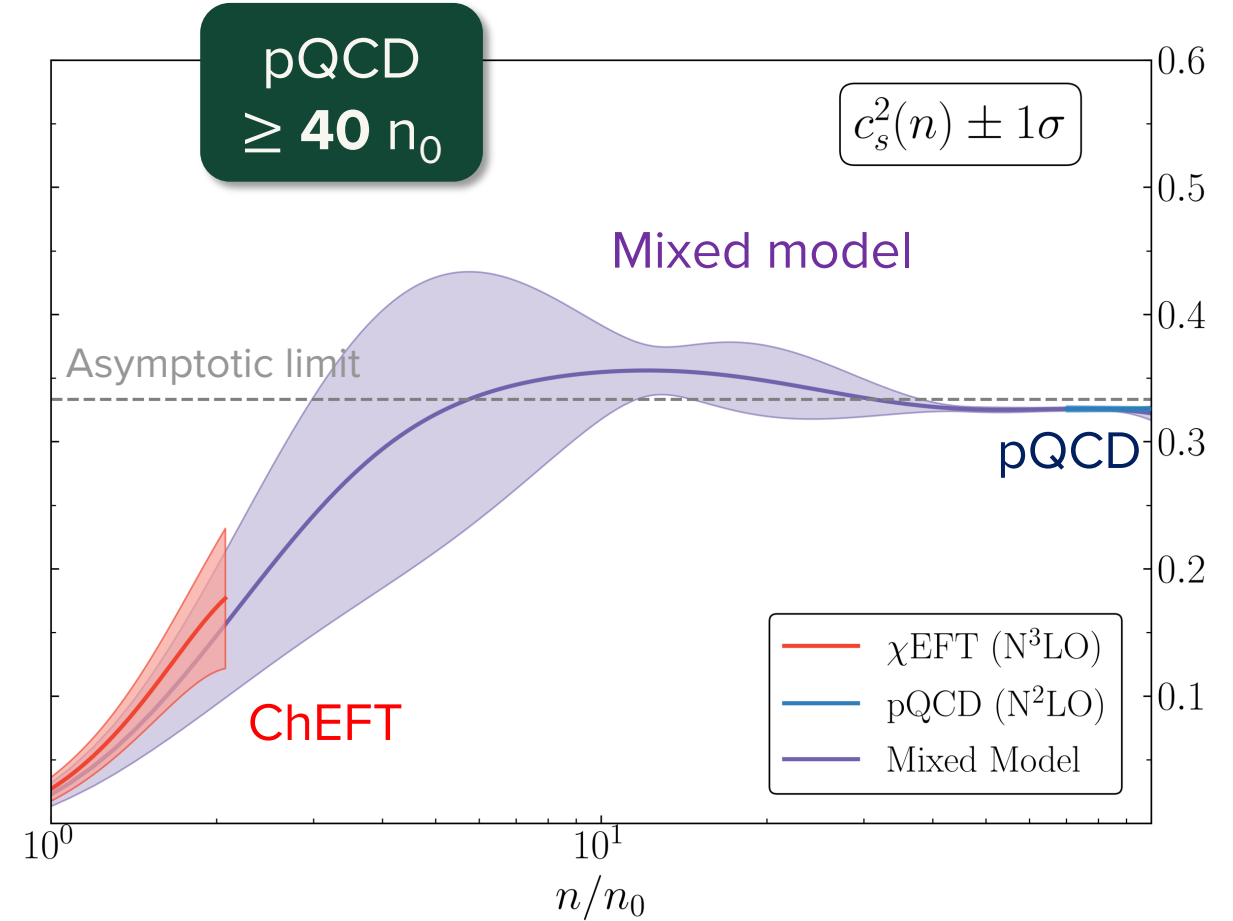
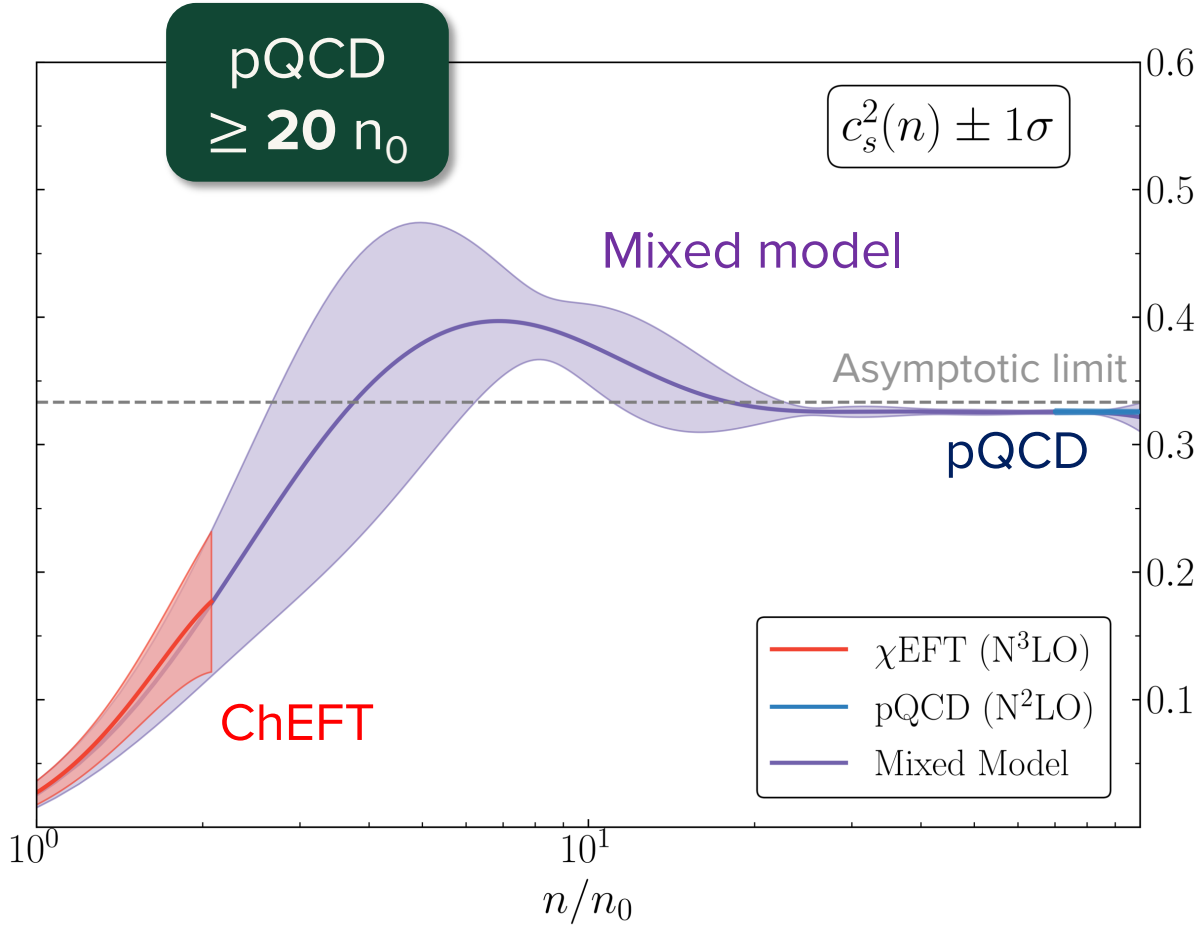
**Constraining the correlation length between chiral EFT & pQCD is crucial to avoid unphysical model correlations at low densities**



# Results: speed of sound

Mixed models both approaching 0.33 asymptotically (from below)

$$c_s^2(n) = \frac{1}{\mu} \frac{\partial P}{\partial n}$$



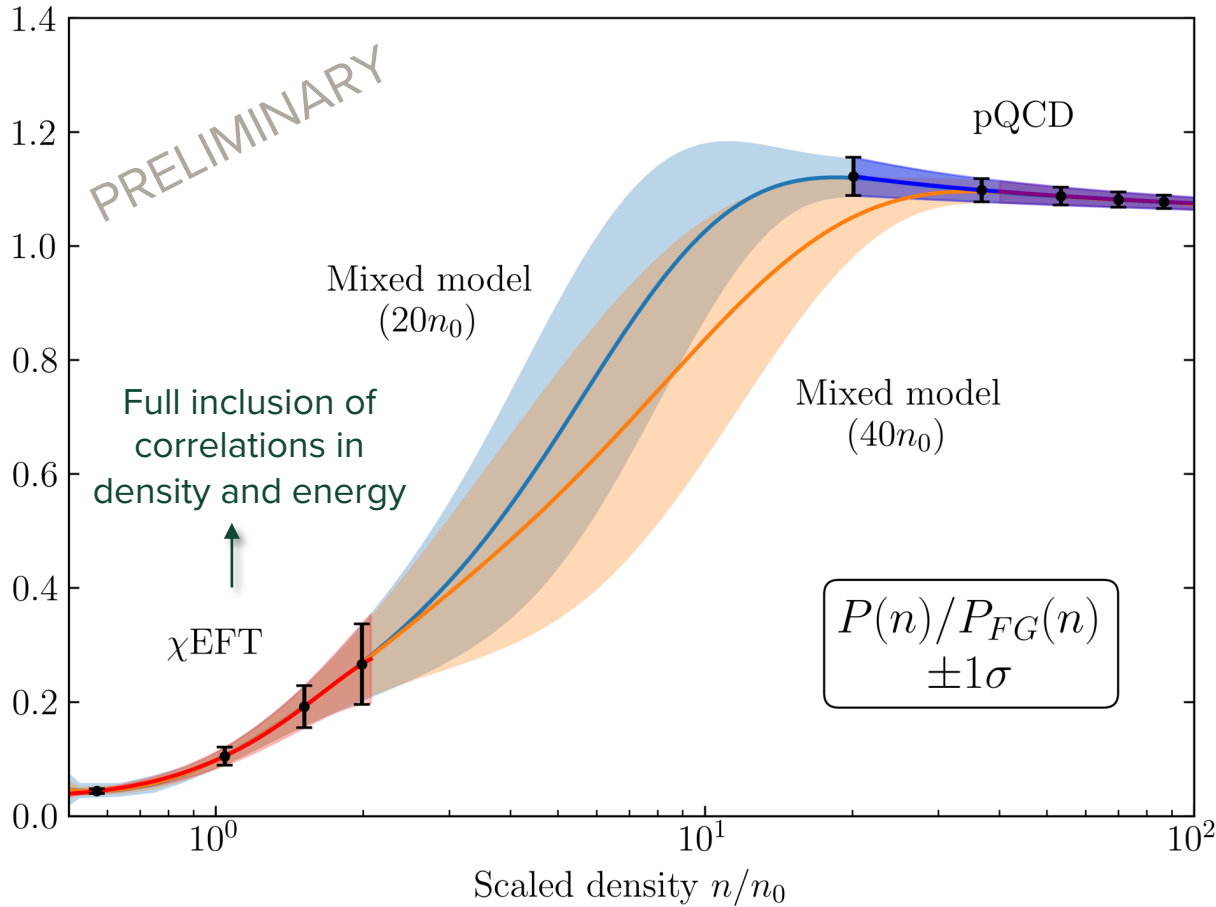
ACS et al., *arXiv:2404.06323*

**Causal, stable EOS**

**GP kernel:** stationary, smooth RBF (as with SNM)  
*[Other stationary kernels give similar results]*

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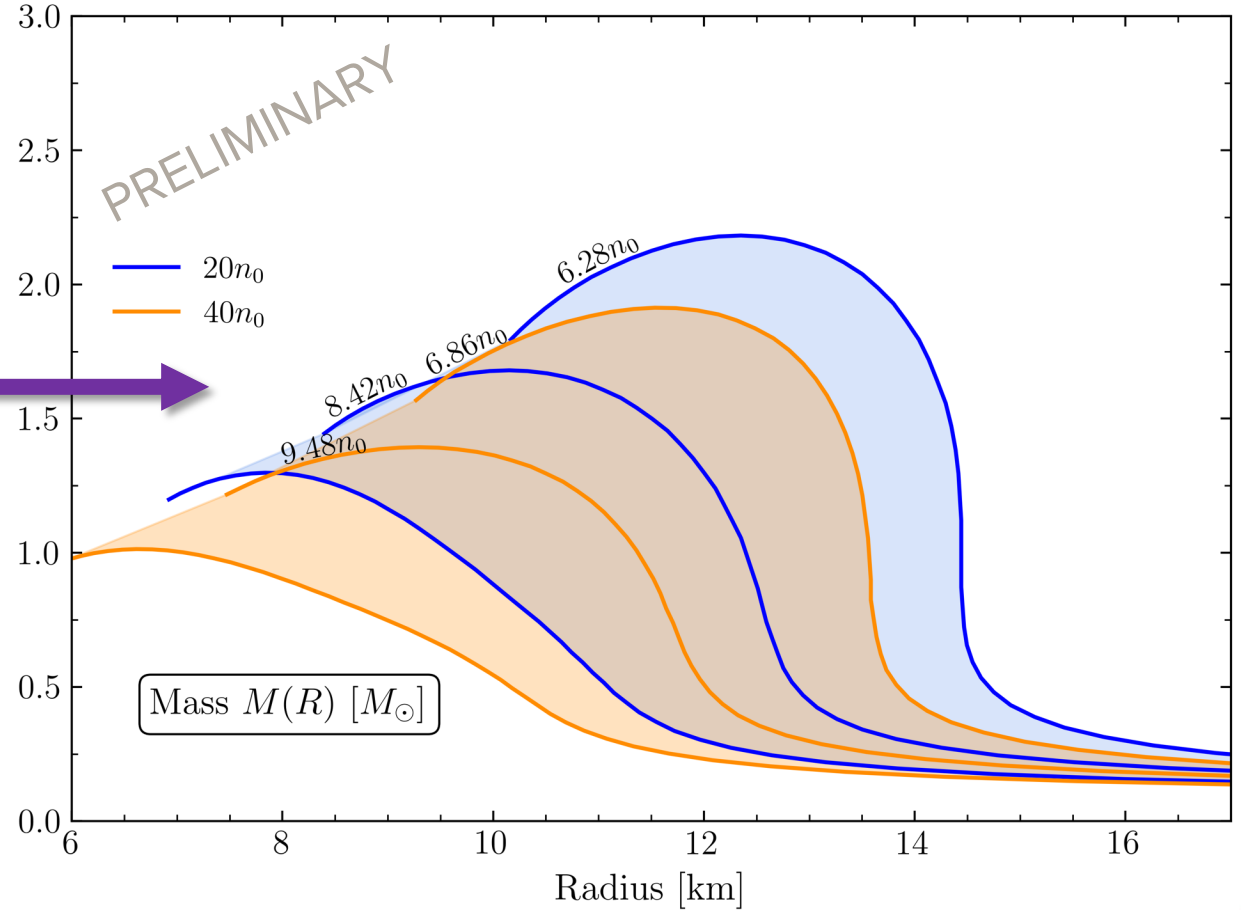
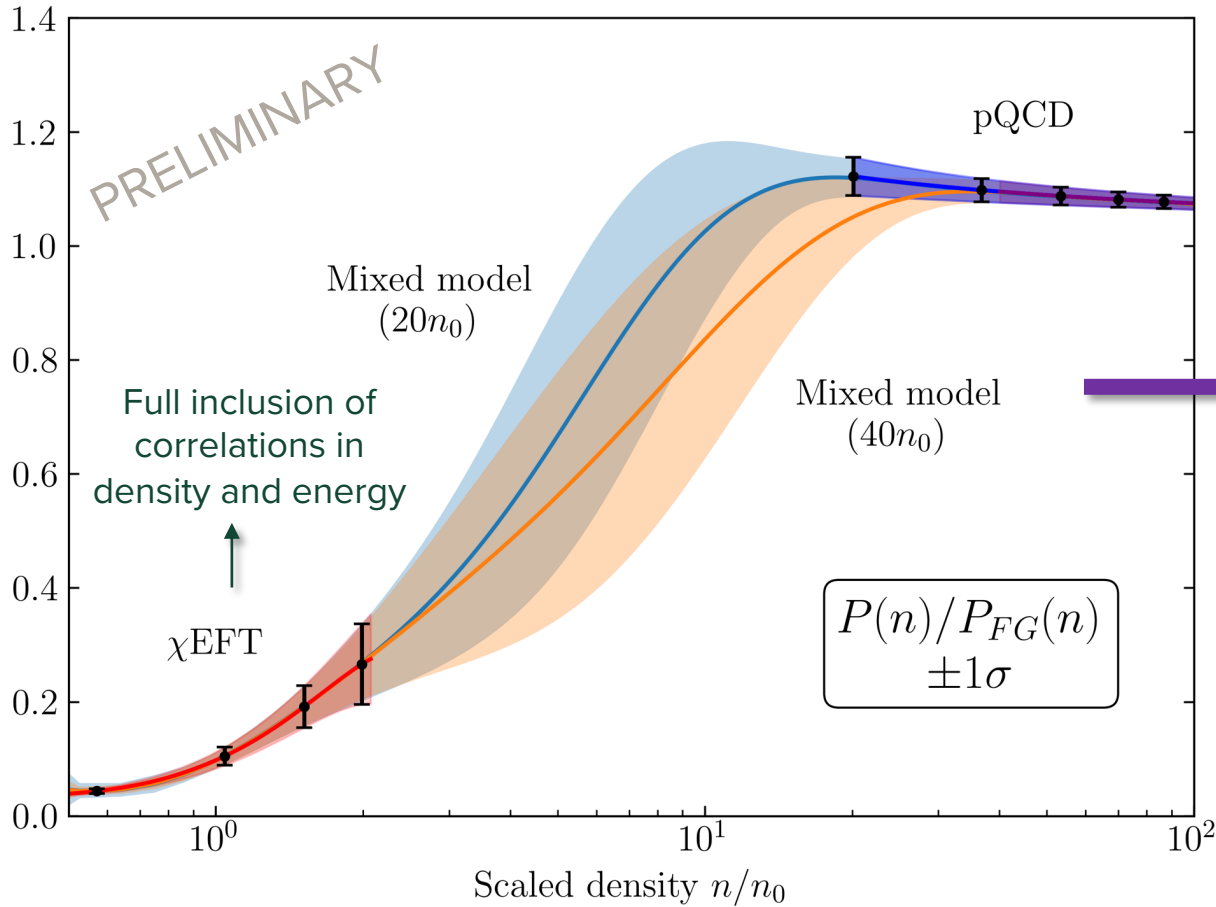
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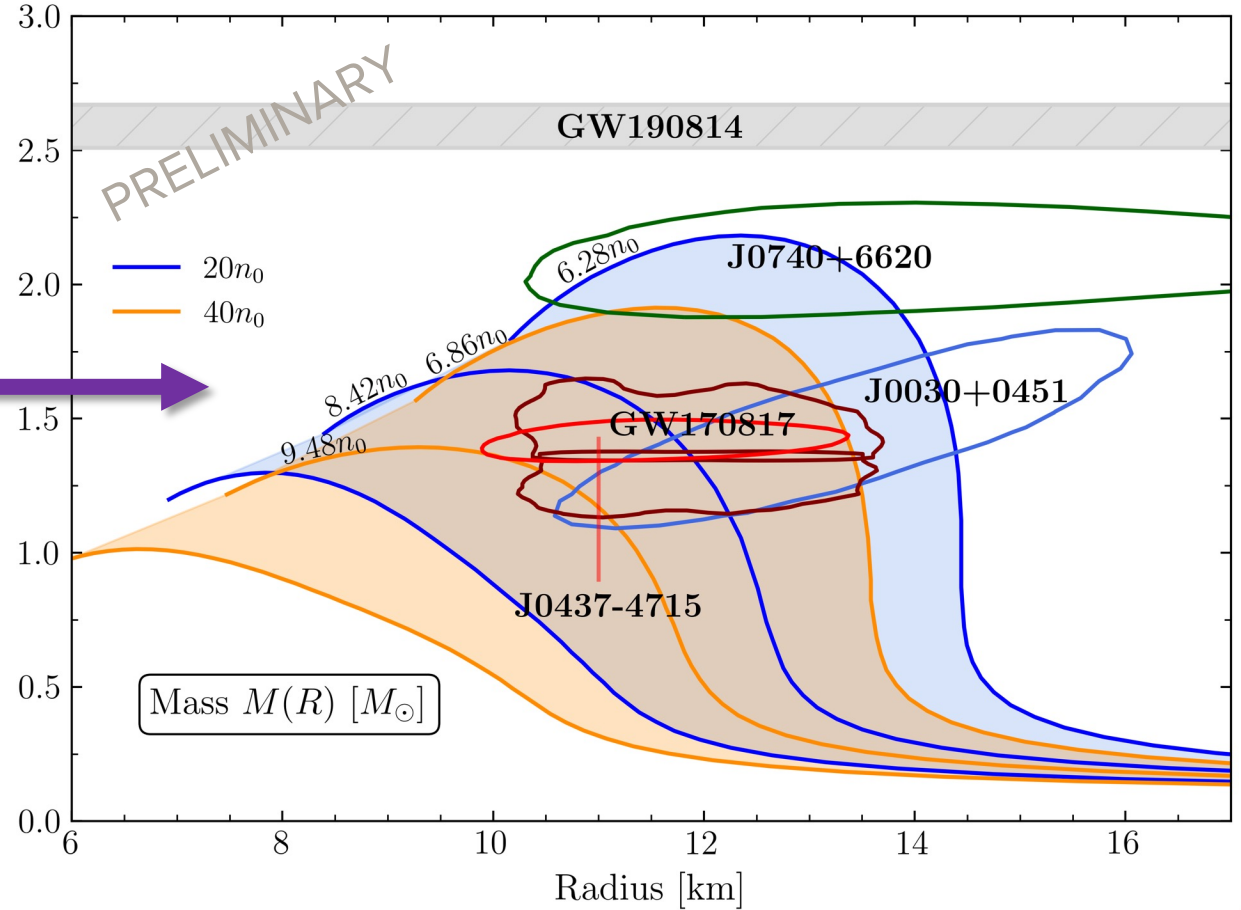
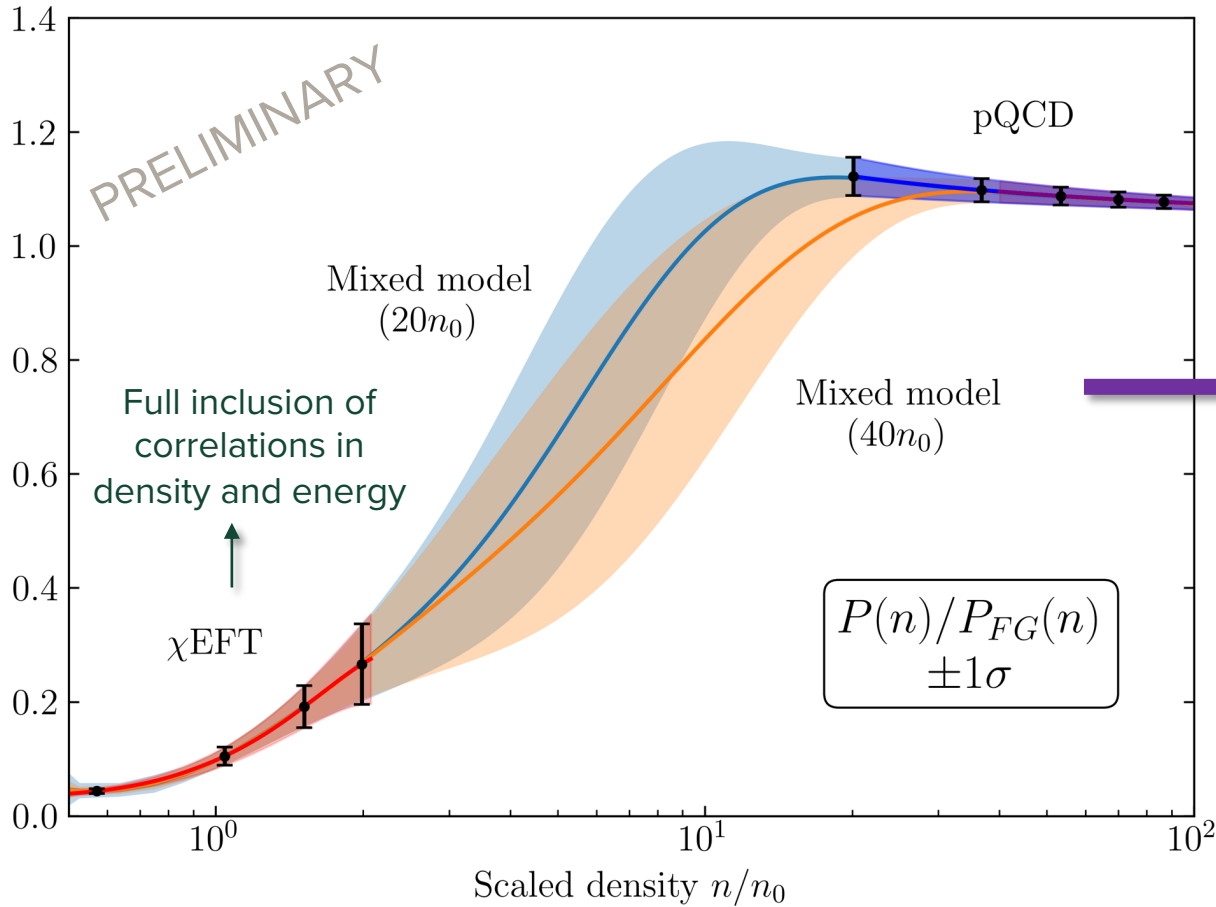
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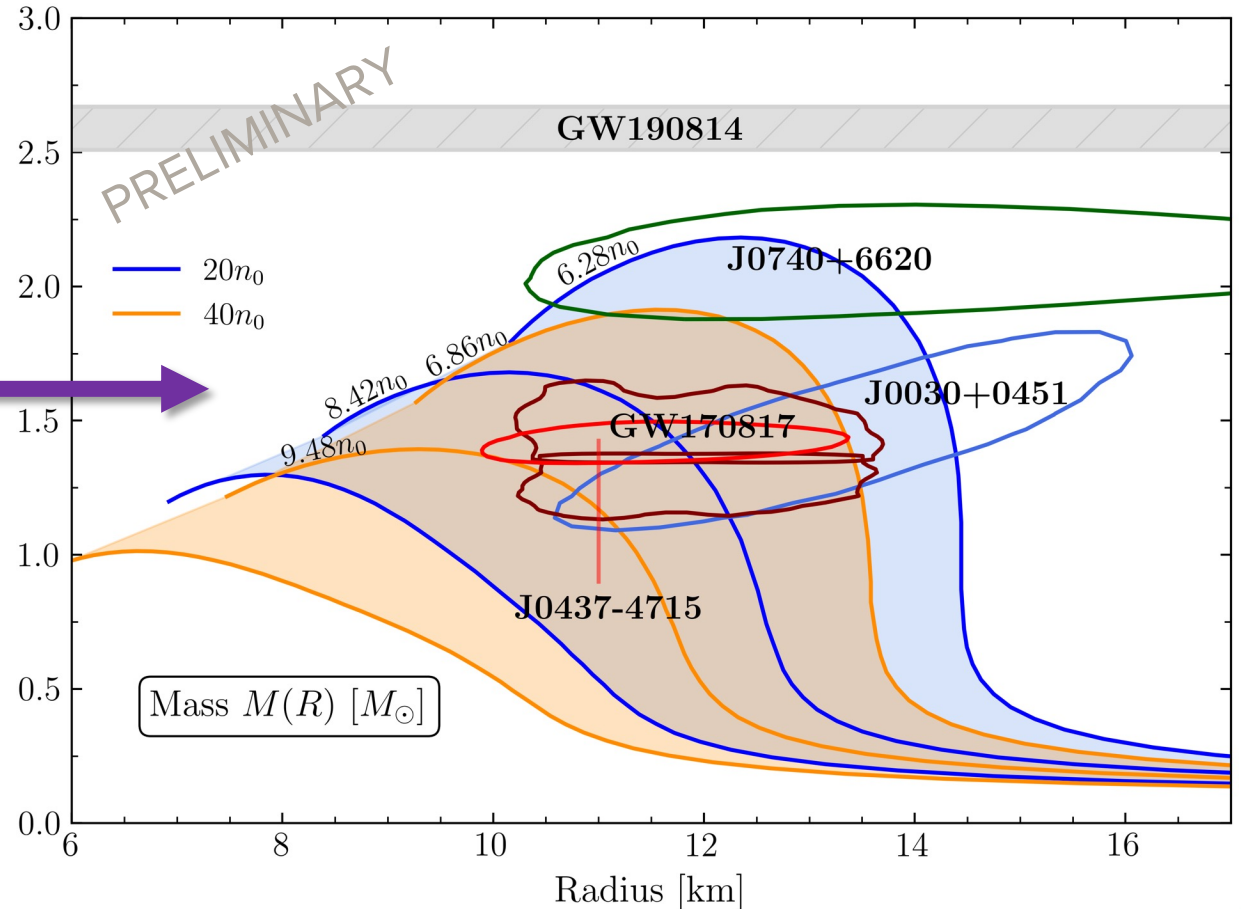
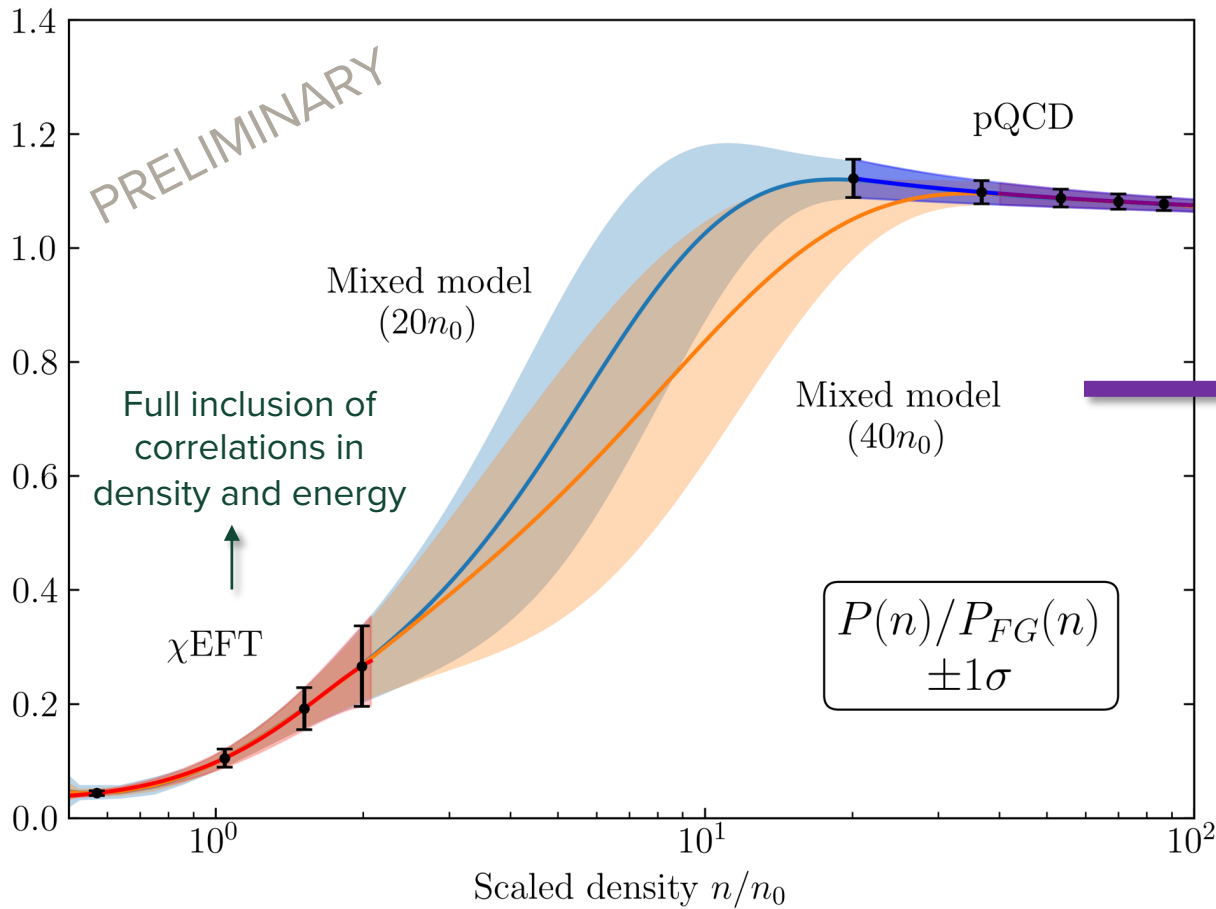
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**Microscopic input for a Bayesian framework**

## Summary

Applied Bayesian model mixing to ChEFT and pQCD in **pressure** for **SNM** and **ANM**

Examined the **speed of sound** results for the mixed EOS

Quantified the **truncation error** in pQCD using BUQEYE methods and consistently obtained **P(n)**

**Goals:** multi-dimensional extension of BMM to finite  $T, \delta$ ; integration with **MUSES** framework

## Current adventures

### Exploring phase transitions

Inclusion of discontinuous phase transitions through non-stationary GP kernels

### Confronting the mixed model with data

Implementing the Bayesian framework to further constrain the posterior of the EOS with astrophysical and heavy-ion results

## ... and stuff we left out

### Including a microscopic crust

Use results from neural-network quantum states (includes clusters), learn crust-core transition

### Full UQ of chiral EFT

Low energy constant (LEC) uncertainties not included, many-body uncertainties may be underestimated



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## Confronting the mixed model with data

Implementing the Bayesian framework to further constrain the posterior of the EOS with astrophysical and heavy-ion results

**Our end goal:** global, microscopic, QCD-based EOSs for merger simulations, but...

## ... and stuff we left out

**Including a microscopic crust**  
Use results from neural-network quantum states (includes clusters), learn crust-core transition

**Full UQ of chiral EFT**  
Low energy constant (LEC) uncertainties not included, many-body uncertainties may be underestimated



**BMM is generally applicable to problems in dense matter**

# Taweret: BMM software

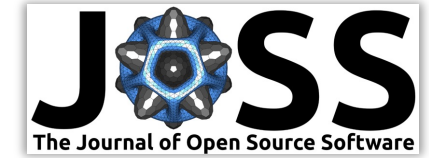
Open-source repository:  
<https://github.com/bandframework/Taweret>

Python interface for user to use BMM methods on **their own models**

Present: our 3 methods + toy models

**Near future:** *add your own BMM method!*

Published in



John Yannotty

Kevin Ingles

Dan Liyanage

Picture credit: Dick Furnstahl

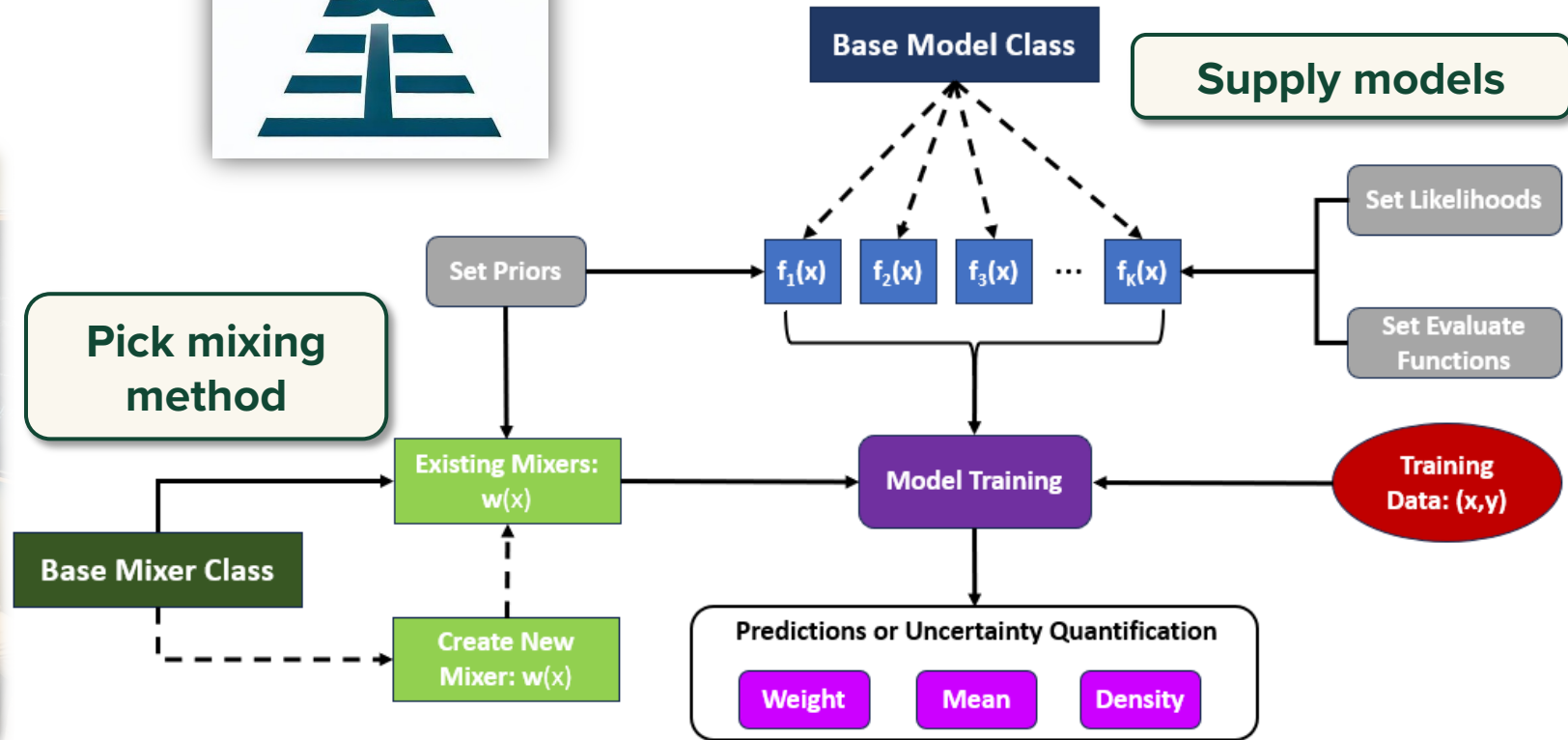


Figure credit: John Yannotty



Daniel Phillips (OU)



Dick Furnstahl (OSU)



Christian  
Drischler (OU)



Jordan  
Melendez

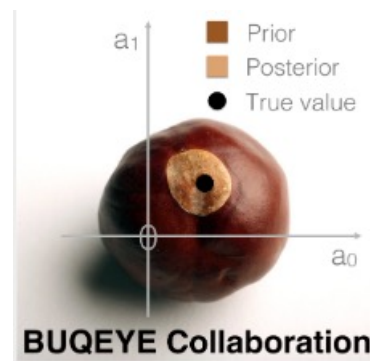
# Thank you!



U.S. DEPARTMENT OF  
**ENERGY**



ASCSN

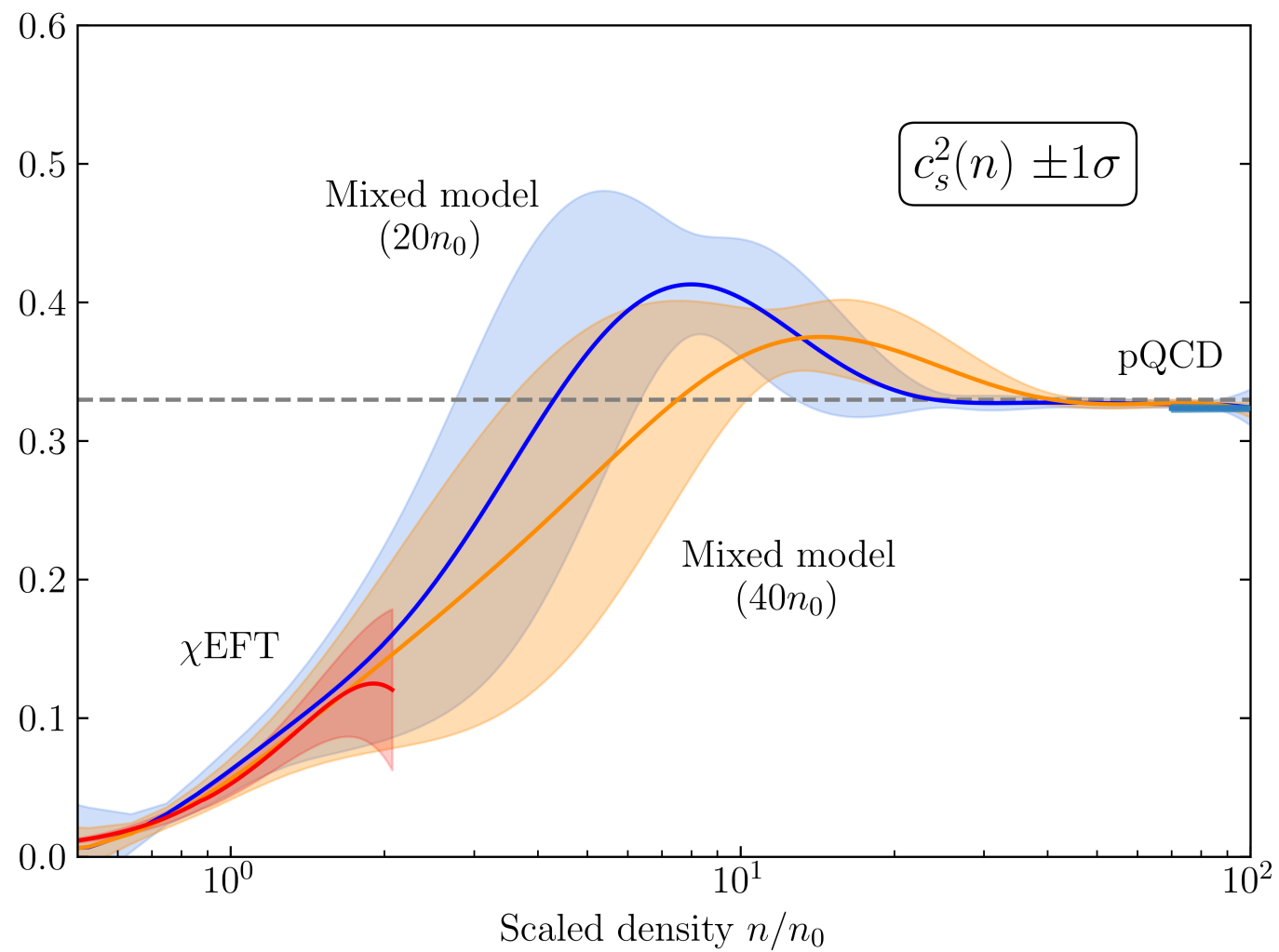


ACS supported by: US DOE, contract DE-FG02-93ER-40756,  
NSF CSSI program, award number OAC-2004601

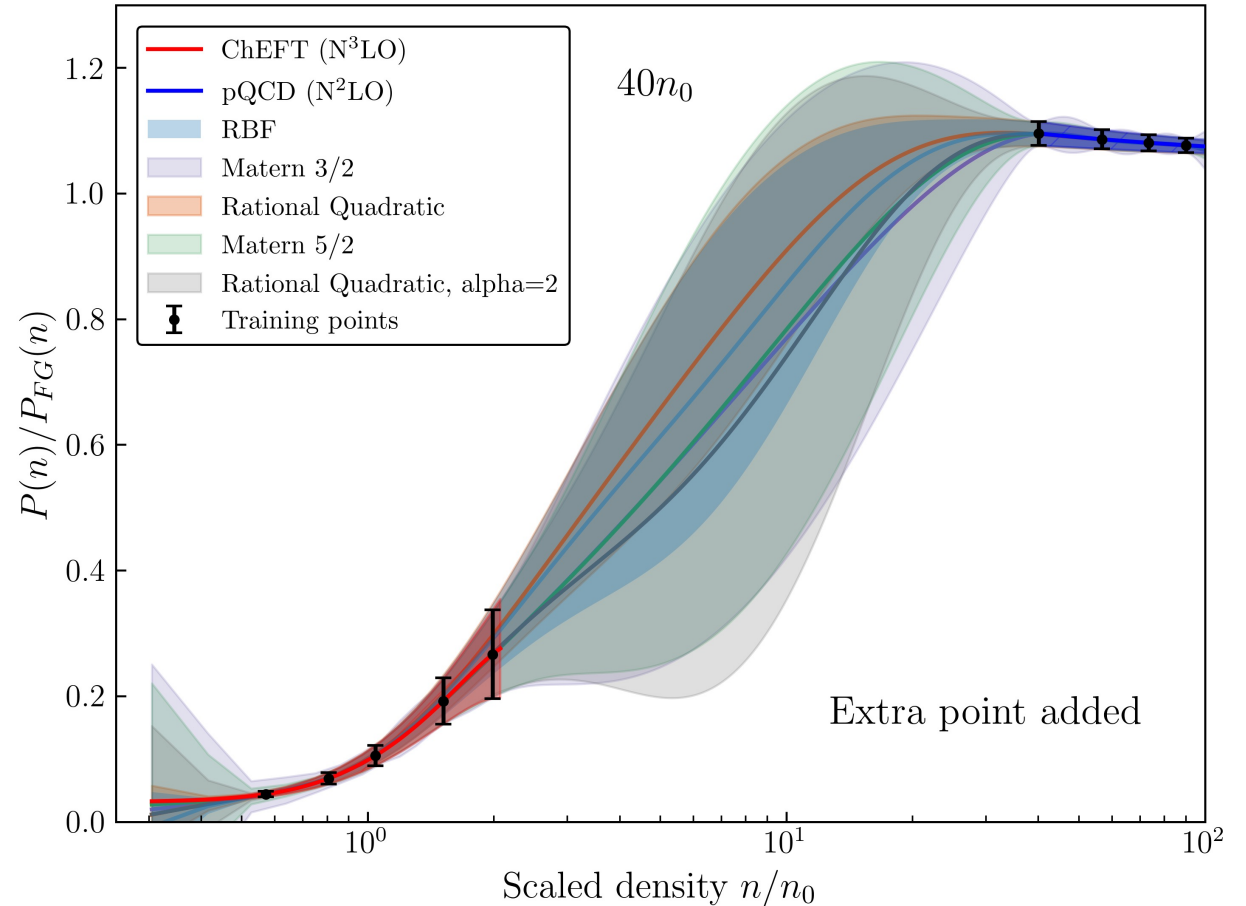
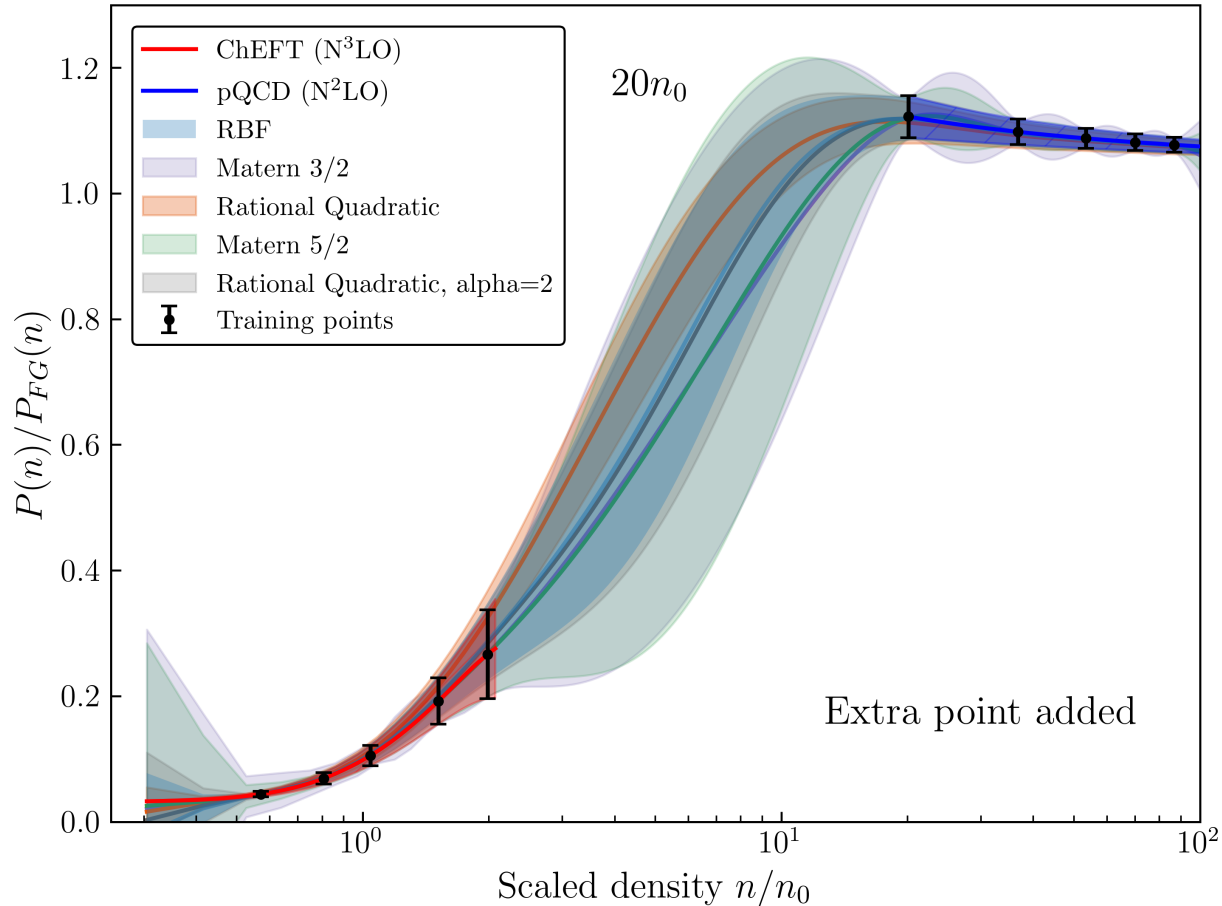


# Backup slides

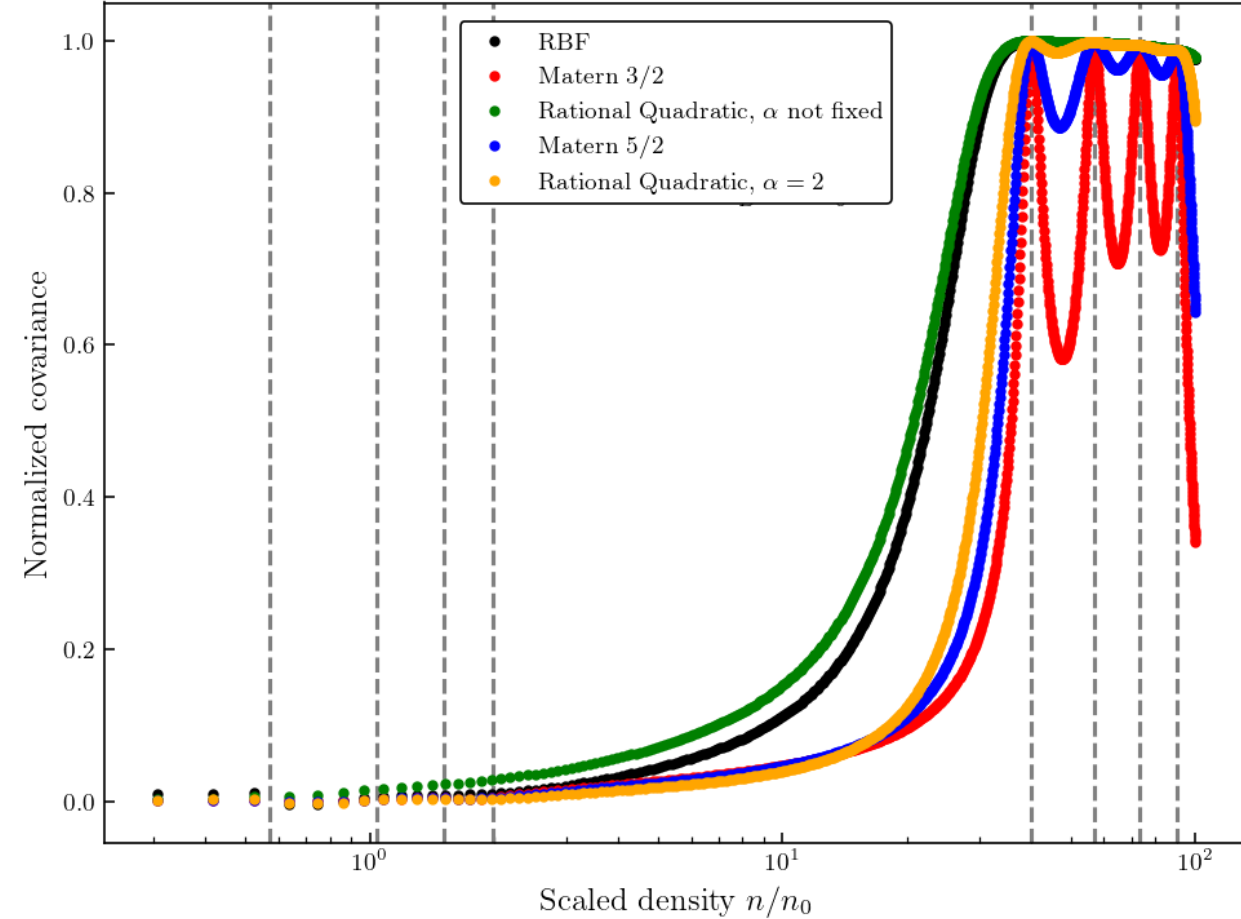
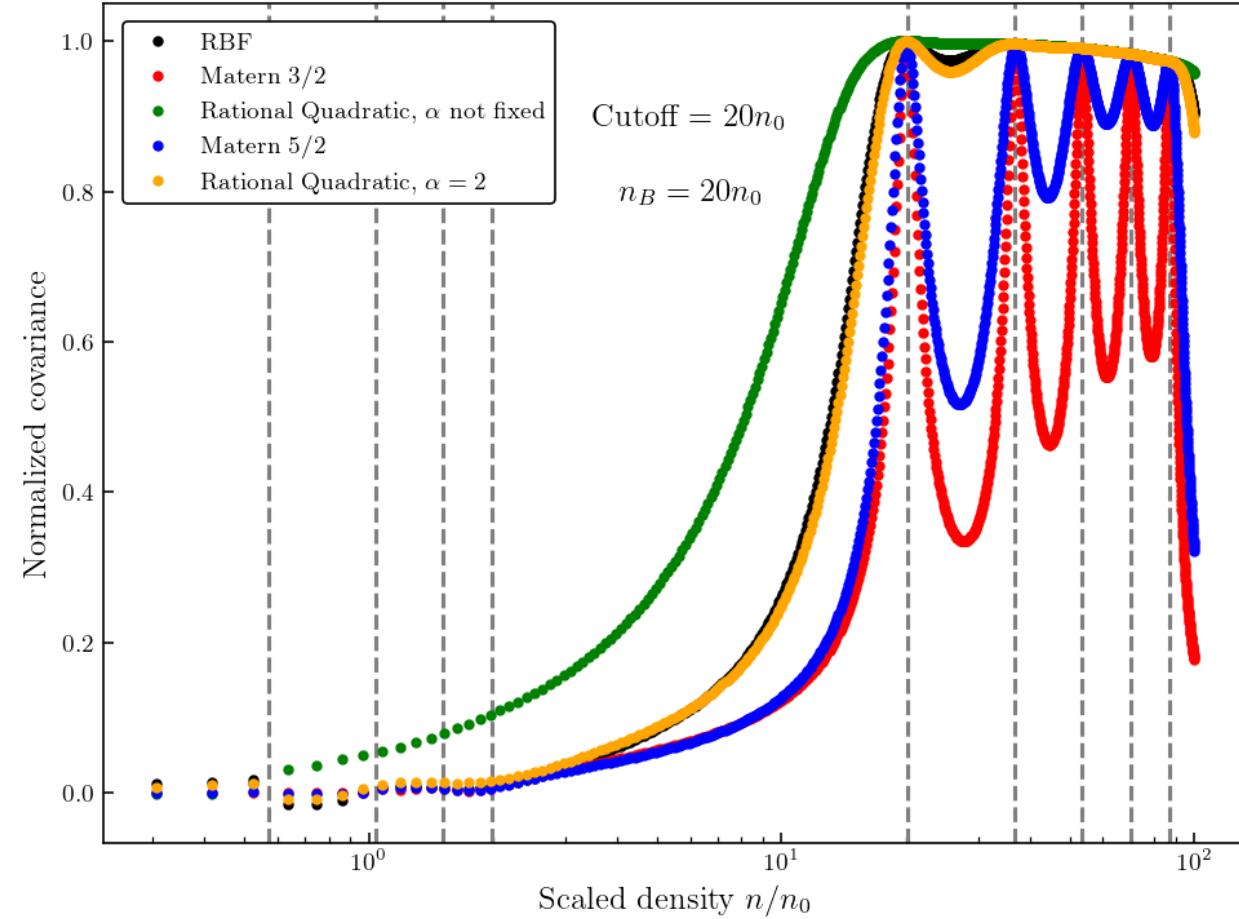
# Speed of sound



# Stationary kernel investigation



# Covariance investigation



# Crash course: Gaussian Processes (GPs)

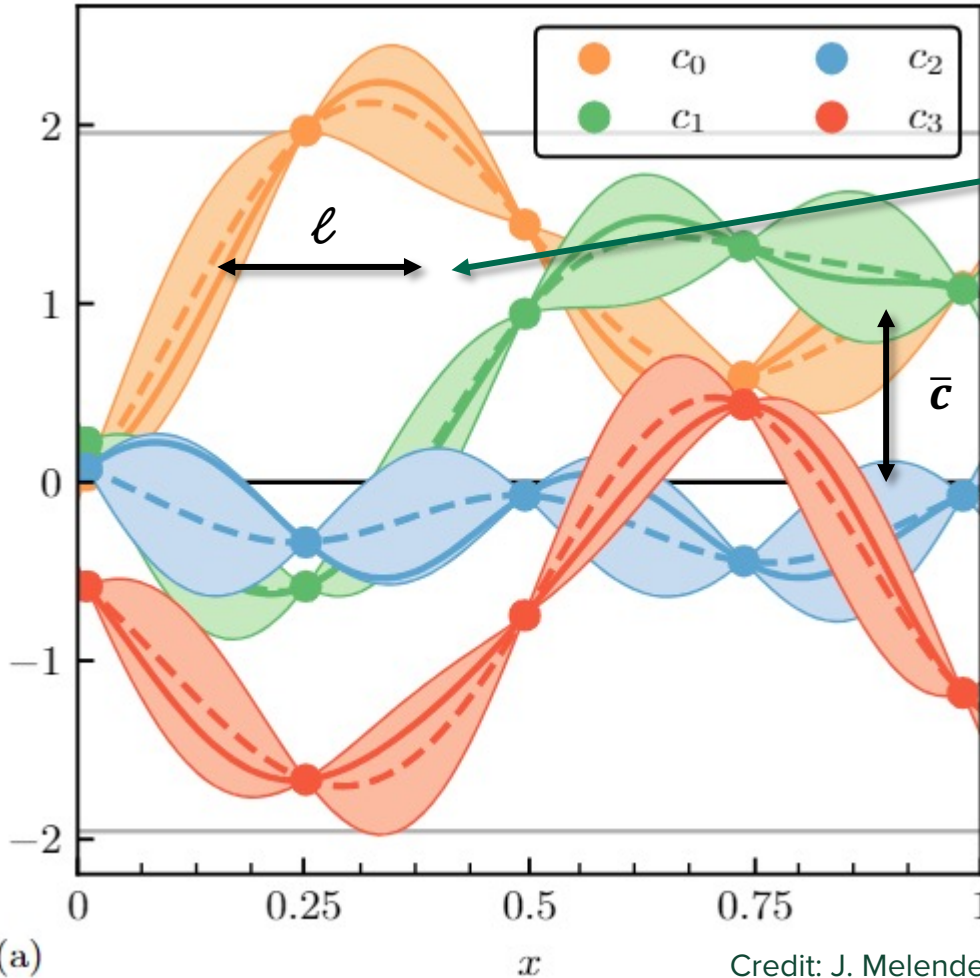
“Set of random variables, any subset of which possesses a Gaussian distribution”

Less abstract:

Defined by mean function and covariance function (*kernel*)

$$f(x) \sim \mathcal{GP}[m(x), \kappa(x, x')]$$

Contains dependence on variance and lengthscale (RBF, Matérn, etc.)



Correlation structure parameterized by a lengthscale and variance

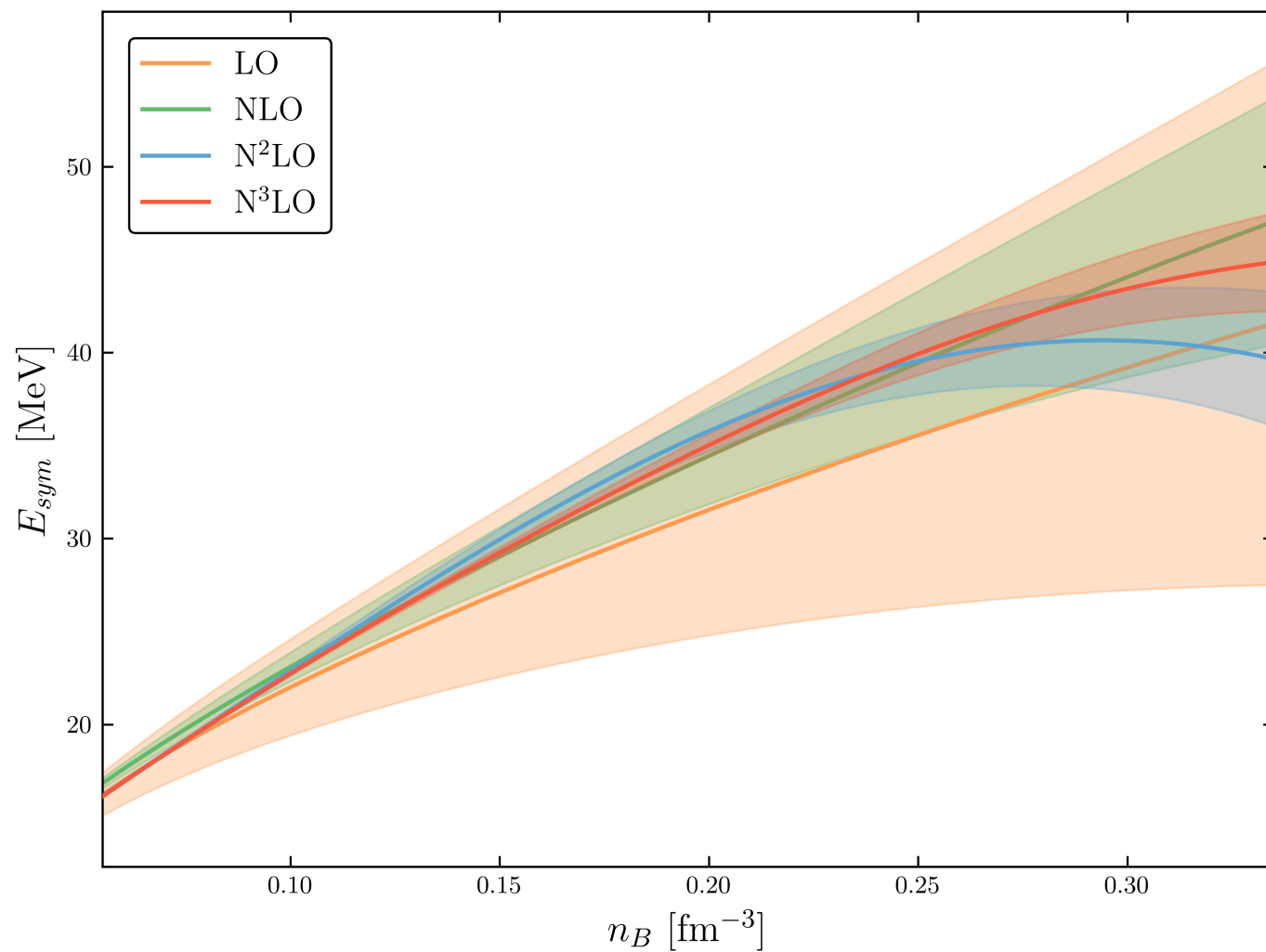
Hyperparameters that *can* be determined by Bayesian parameter estimation

Can be fitted to **data + uncertainties** and used to predict at new points

Credit: J. Melendez et al. (2019)



# Symmetry energy from chiral EFT



# High densities: pQCD EOS

Invoking the Kohn-Luttinger-Ward inversion theorem

$$P(\mu) = P_{FG}(\mu) \left[ c_0 + c_1 Q(\bar{\Lambda}) + c_2(\mu) Q^2(\bar{\Lambda}) \right]$$

**Goal:**  $P(\mu) \rightarrow P(n)$

**1**  $\mu = \mu_{FG} + \mu_1 + \mu_2$  ← Perturbative expansion

**2** Taylor expand  $\Rightarrow n(\mu) = \frac{\partial P(\mu)}{\partial \mu} \rightarrow n(\mu_{FG}) \equiv n(\mu_{FG} + \mu_1 + \mu_2) \equiv \bar{n}$  Input number density

**3** Equate terms by counting powers of  $\alpha_s$

$$\bar{n}(\mu) = c_0(\mu) \frac{\partial P_{FG}(\mu)}{\partial \mu} \Big|_{\mu=\mu_{FG}}, \quad \mu_1 = - \frac{c_1 Q(\bar{\Lambda}) \frac{\partial P_{FG}(\mu)}{\partial \mu}}{c_0 \frac{\partial^2 P_{FG}(\mu)}{\partial \mu^2}} \Big|_{\mu=\mu_{FG}} + \mu_2 \text{ expression}$$

**4** Expand  $P(\mu)$ , insert terms, keep up to second order in  $\alpha_s$

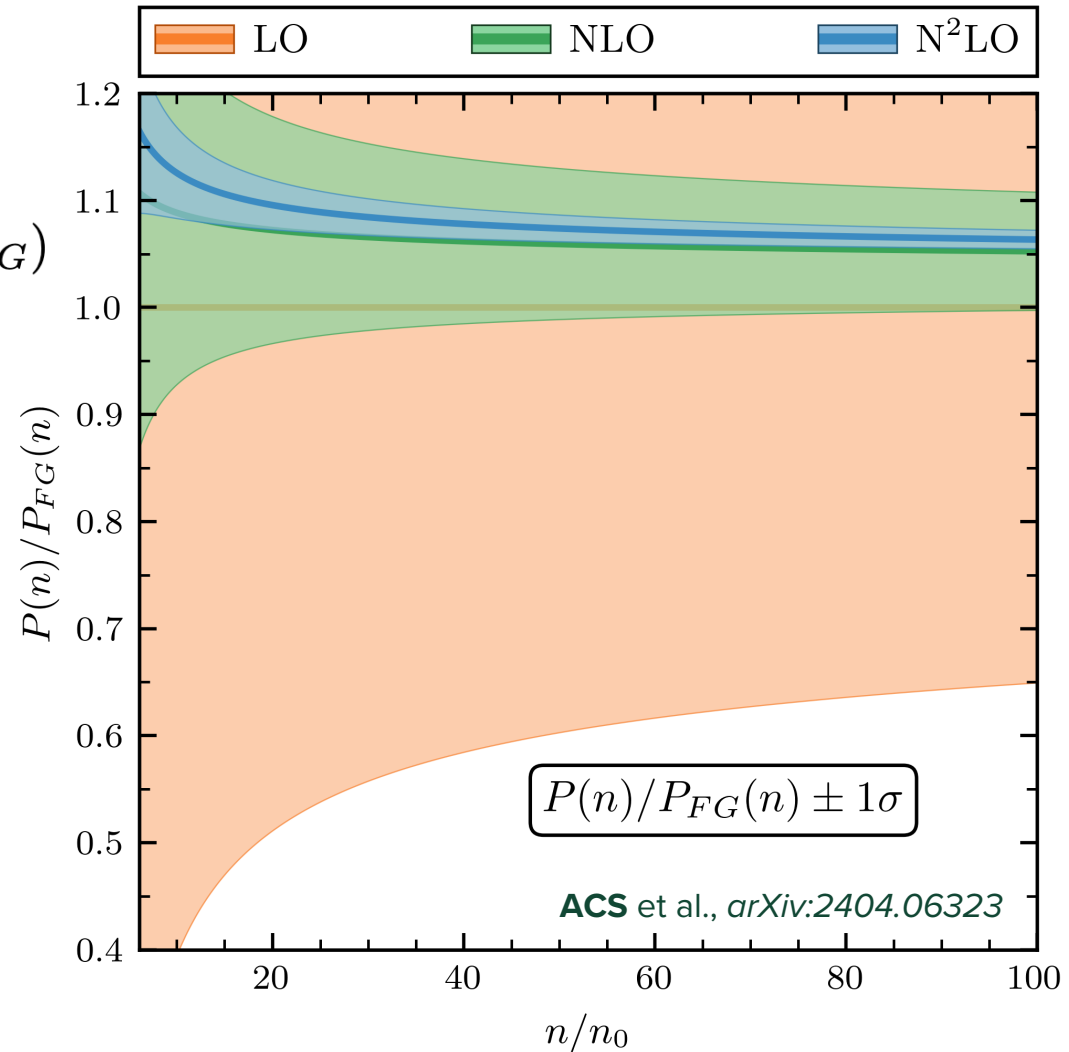
$$\frac{P(n)}{P_{FG}(n)} = 1 + \frac{2}{3\pi} \alpha_s(\bar{\Lambda}_{FG}) + \frac{8}{9\pi^2} \alpha_s^2(\bar{\Lambda}_{FG}) - \frac{N_f^2}{3\pi^2} c_2(\mu_{FG}) \alpha_s^2(\bar{\Lambda}_{FG}) - \frac{\beta_0}{3\pi^2} \alpha_s^2(\bar{\Lambda}_{FG})$$

Directly  $n$ -dependent

# High densities: pQCD EOS

Expansion in the strong coupling constant  $\alpha_s$

$$P(n) = P_{FG}(n) \left[ c_0 + c_1 Q(\bar{\Lambda}_{FG}) + c_2(n) Q^2(\bar{\Lambda}_{FG}) + \dots \right]$$



# High densities: pQCD EOS

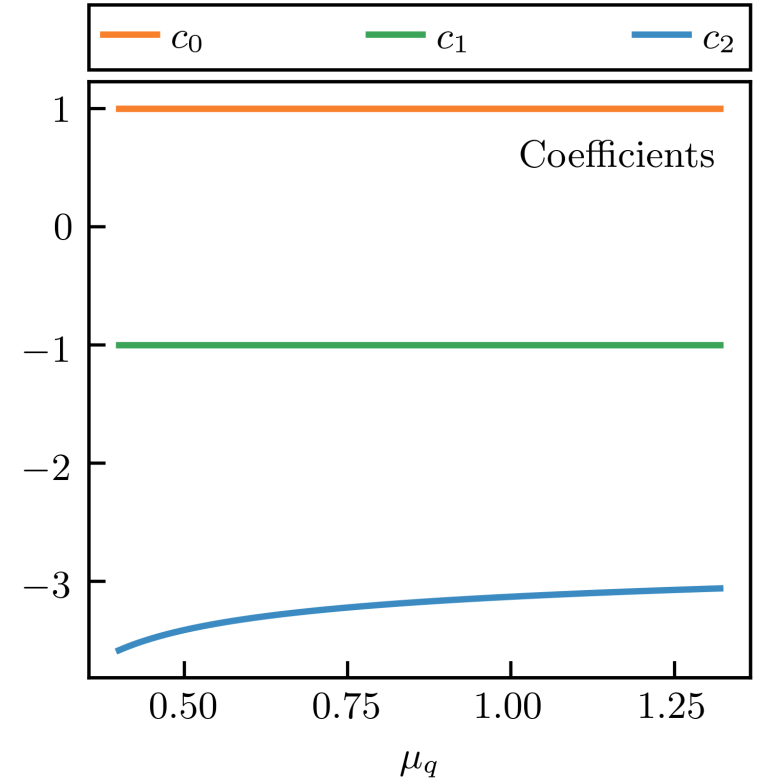
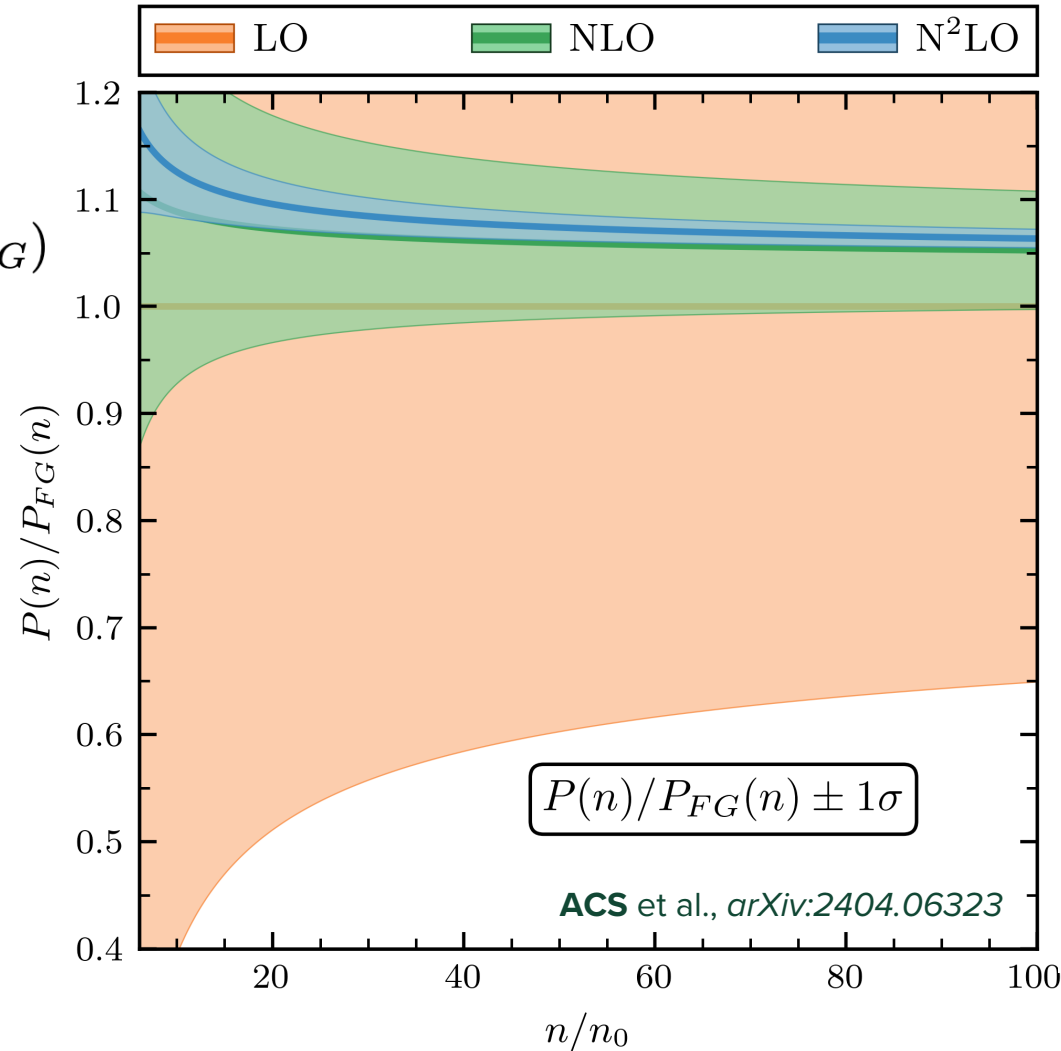
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Apply **gsum** to **P(n)**

$$Q = \frac{N_f}{\pi} \alpha_s(\bar{\Lambda})$$

$$y_{\text{ref}} = P_{FG}(n)$$



**Truncation error** often assessed by varying renormalization scale to obtain band---not as statistically rigorous

\* See newer works by Gorda et al. (2022, 2023), MiHO model for higher orders



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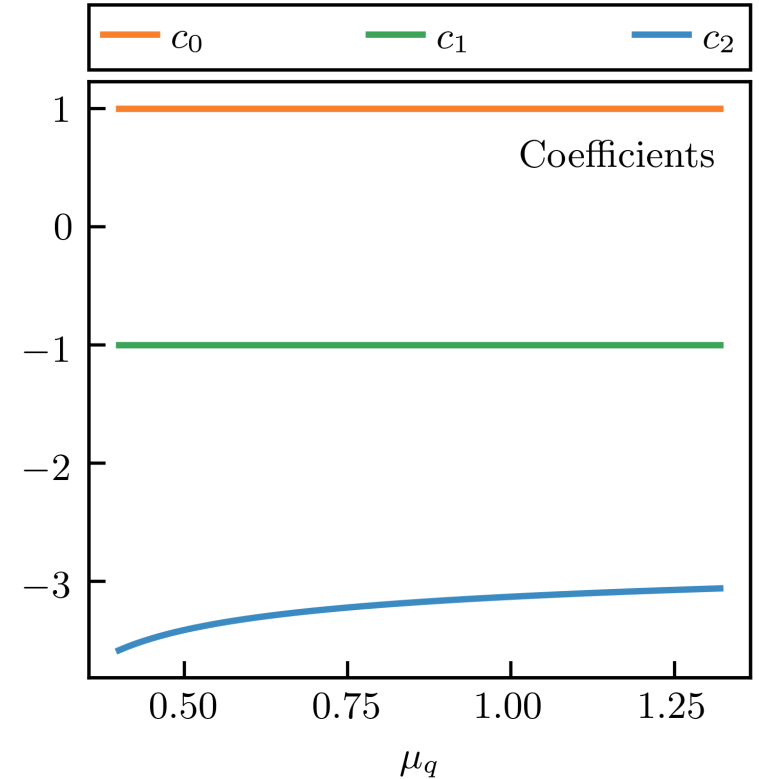
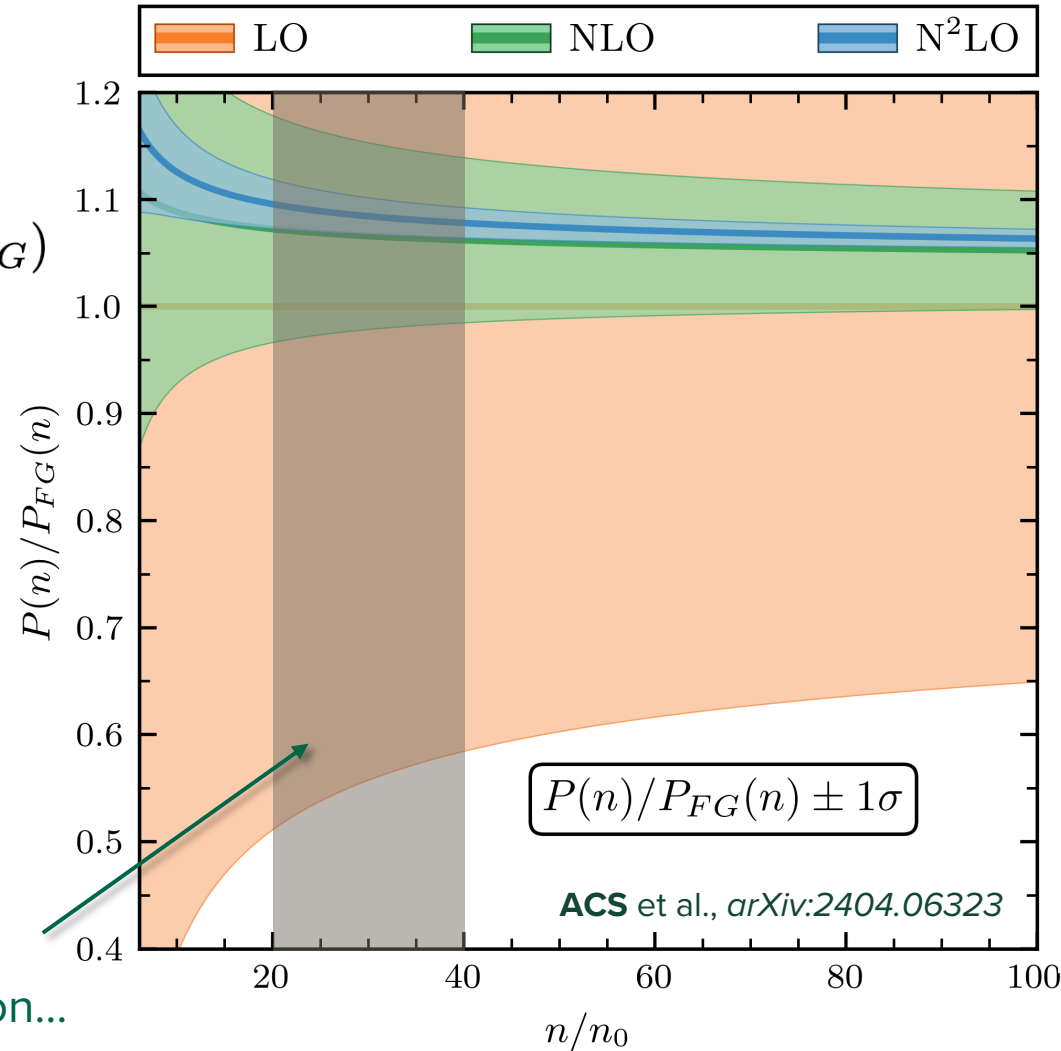
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Nonperturbative effects under (20-40)  $n_0$ : pairing, hadronization...



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# High densities: pQCD EOS

Calculate **speed of sound squared**:

$$c_s^2(n) = n \frac{\partial \ln(\mu(n))}{\partial n}$$

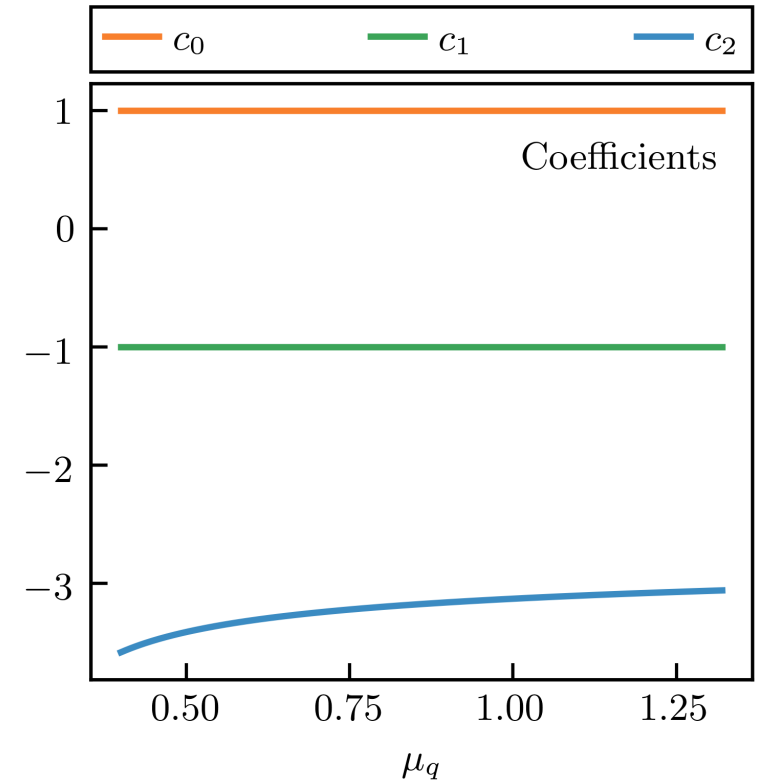
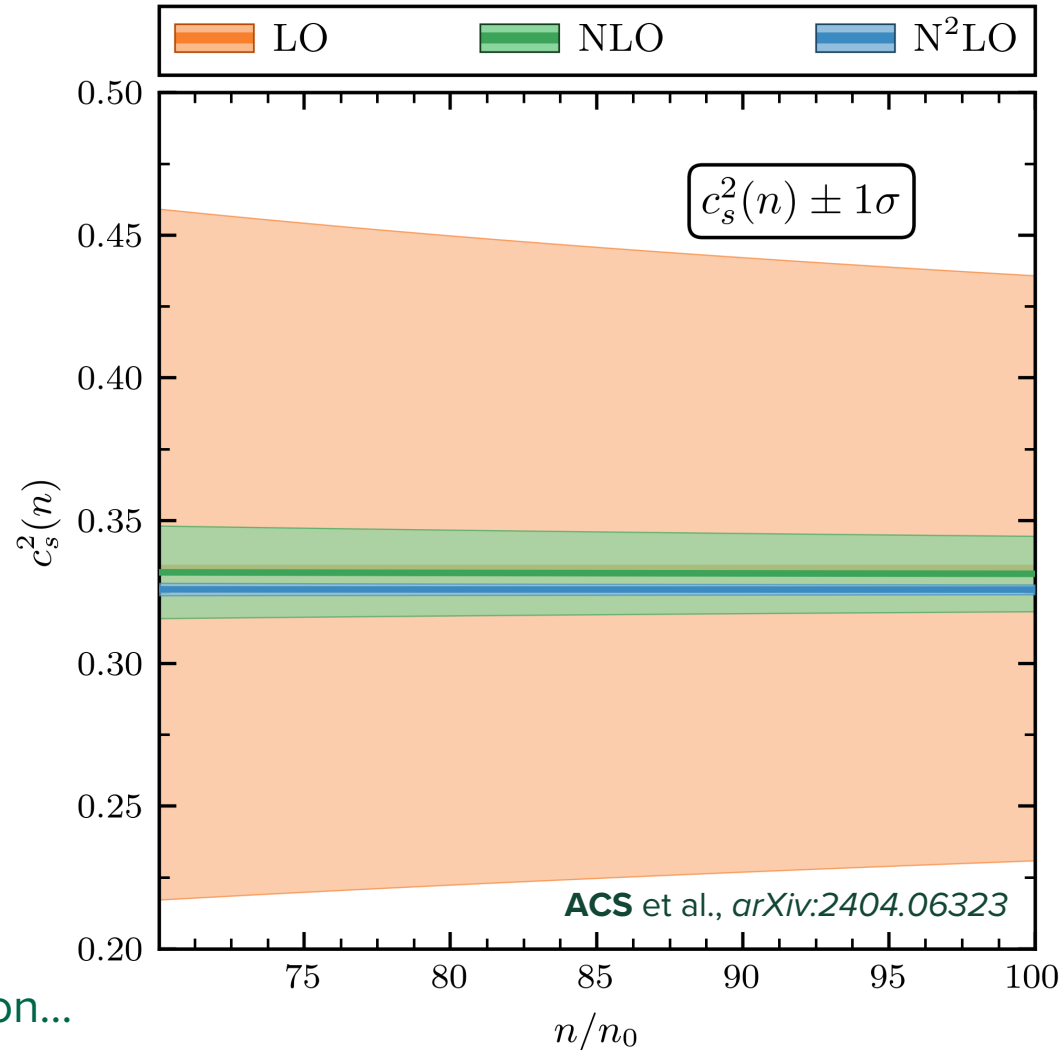
[**Equal** at N<sup>2</sup>LO to the calculation of  $c_s^2(\mu)$  using the pressure directly]

$$Q = \frac{N_f}{\pi} \alpha_s(\bar{\Lambda})$$

$$y_{\text{ref}} = P_{FG}(n)$$



Nonperturbative effects under (20-40)  $n_0$ : pairing, hadronization...



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# collaboration: efforts

v0.4 coming soon!



<https://github.com/bandframework/bandframework>

## Cyber-infrastructure

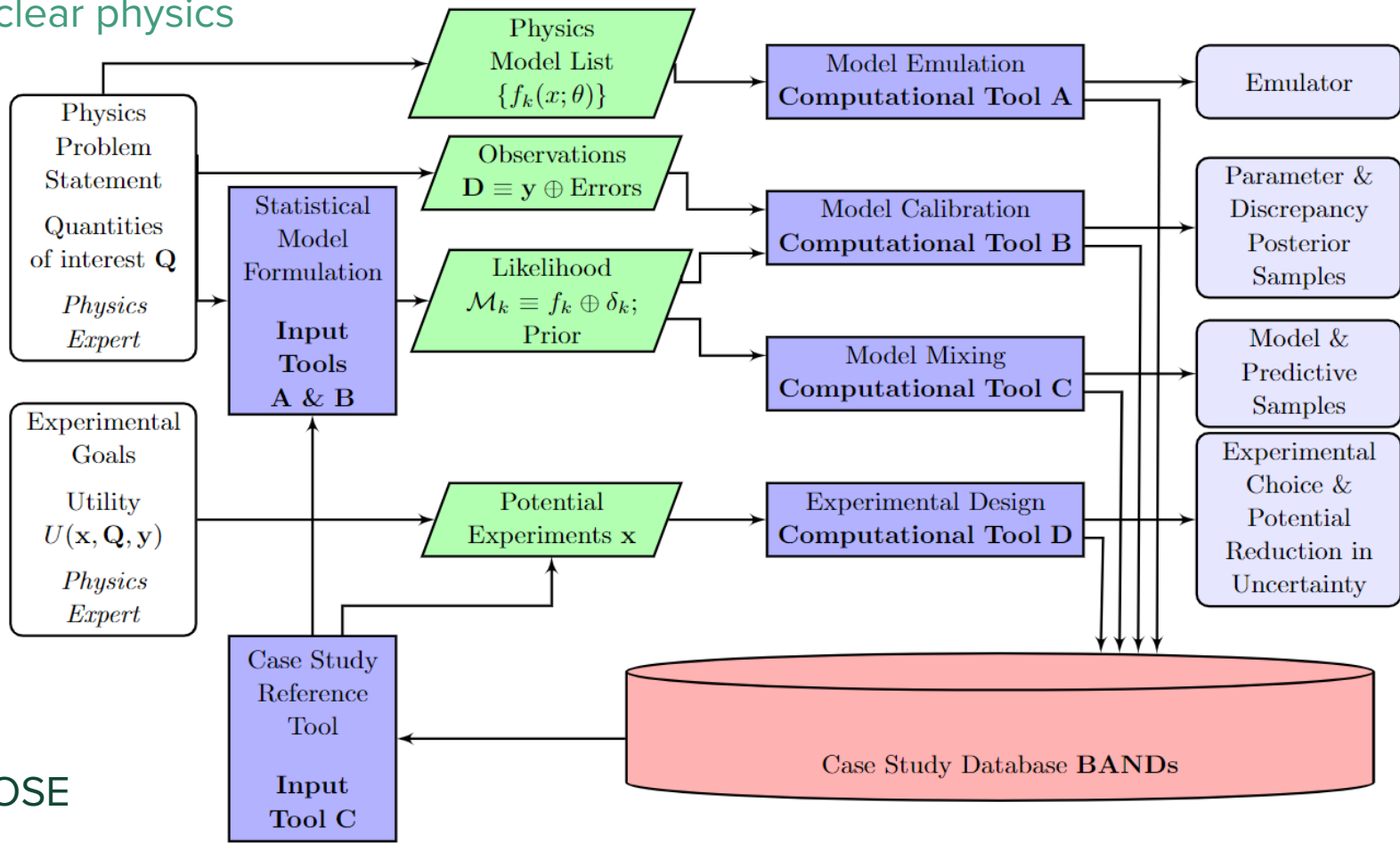
Develop tools for nuclear physics community  
Test on toy models and real-world applications

ParMOO

BMEX

surmise

ROSE



Our model mixing software!

SAMBA



Taweret



# Truncation error: gsum and GPs

GP: governed by mean function and covariance function

$$\delta y_k(x) | \theta, Q \sim \mathcal{GP}[m_{\delta k}(x), \bar{c}^2 R_{\delta k}(x, x'; \ell)] \leftarrow \begin{array}{l} c_n(x) | \theta \stackrel{\text{iid}}{\sim} \mathcal{GP}[\mu, \bar{c}^2 r(x, x'; \ell)], \\ \theta \equiv \{\mu, \bar{c}^2, \ell\}. \end{array} \rightarrow \begin{array}{l} m_{\delta k}(x) \equiv y_{\text{ref}}(x) \frac{Q(x)^{k+1}}{1 - Q(x)} \mu \equiv b_{\delta k}(x) \mu \\ R_{\delta k}(x, x'; \ell) \equiv y_{\text{ref}}(x) y_{\text{ref}}(x') \frac{[Q(x)Q(x')]^{k+1}}{1 - Q(x)Q(x')} r(x, x'; \ell) \end{array}$$

Hyperparameter estimation

Results insensitive to exact form, so use conjugate priors  $\mu, \bar{c}^2 \sim N\chi^{-2}(\eta_0, V_0, \nu_0, \tau_0^2)$

Diagnostic tests for assessing the calibration of the GP

Squared Mahalanobis distance: multi-dimensional sum of squared residuals

$$D_{\text{MD}}^2(\mathbf{f}_{\text{val}}) = (\mathbf{f}_{\text{val}} - \mathbf{m})^\top K^{-1}(\mathbf{f}_{\text{val}} - \mathbf{m})$$

Pivoted Cholesky decomposition: pinpoints data that is yielding a failing MD

$$K = GG^\top \quad \longrightarrow \quad \mathbf{D}_G = G^{-1}(\mathbf{f}_{\text{val}} - \mathbf{m})$$

Credible interval diagnostic: testing accuracy of the emulator

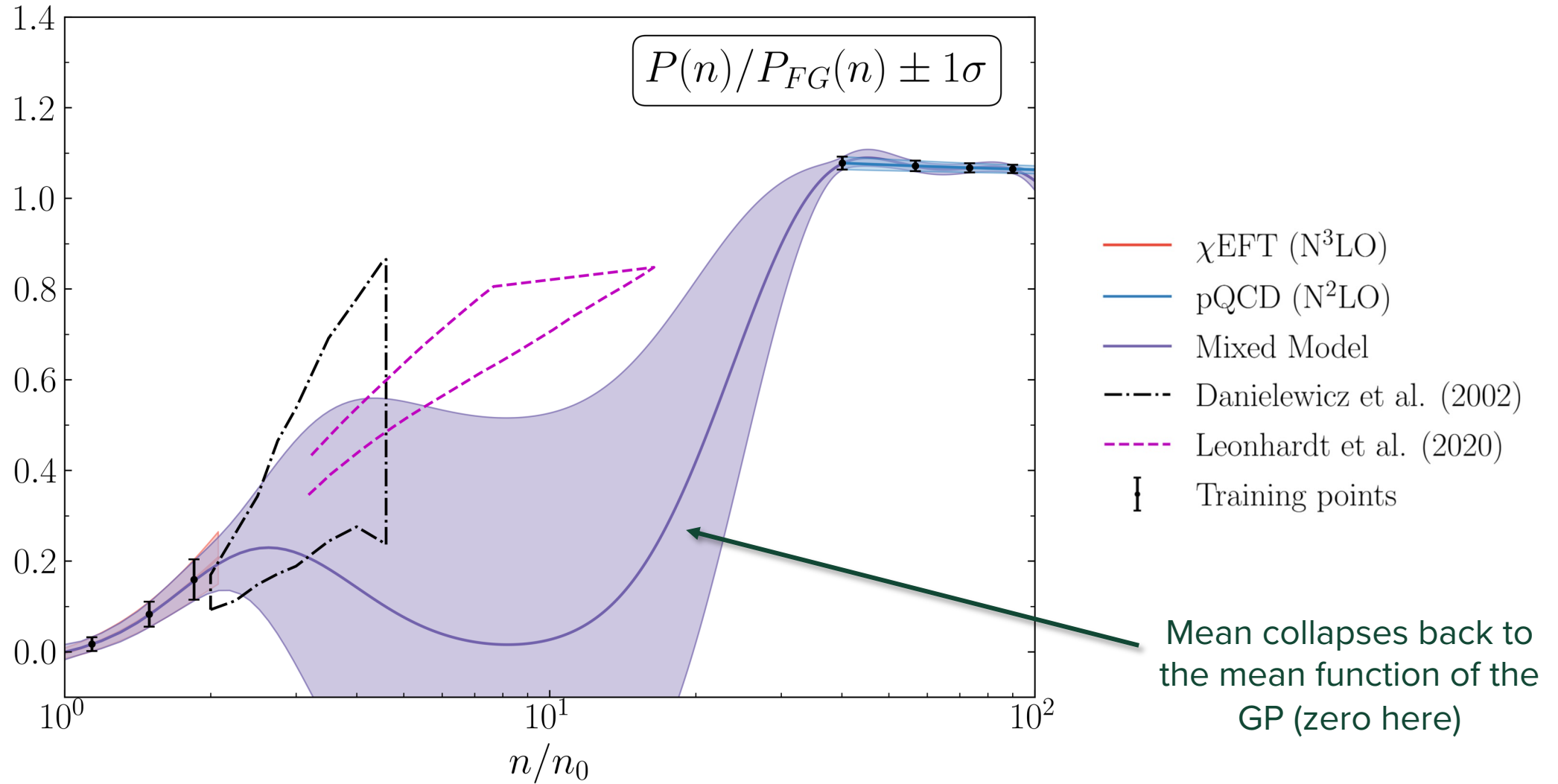
$$D_{\text{CI}}(\alpha, \mathbf{f}_{\text{val}}) = \frac{1}{M} \sum_{i=1}^M \mathbf{1}[f_i \in \text{CI}_i(\alpha)]$$

All equations from J. Melendez et al. (2020)





# Extreme case: very short lengthscale



# Results: pointwise approach

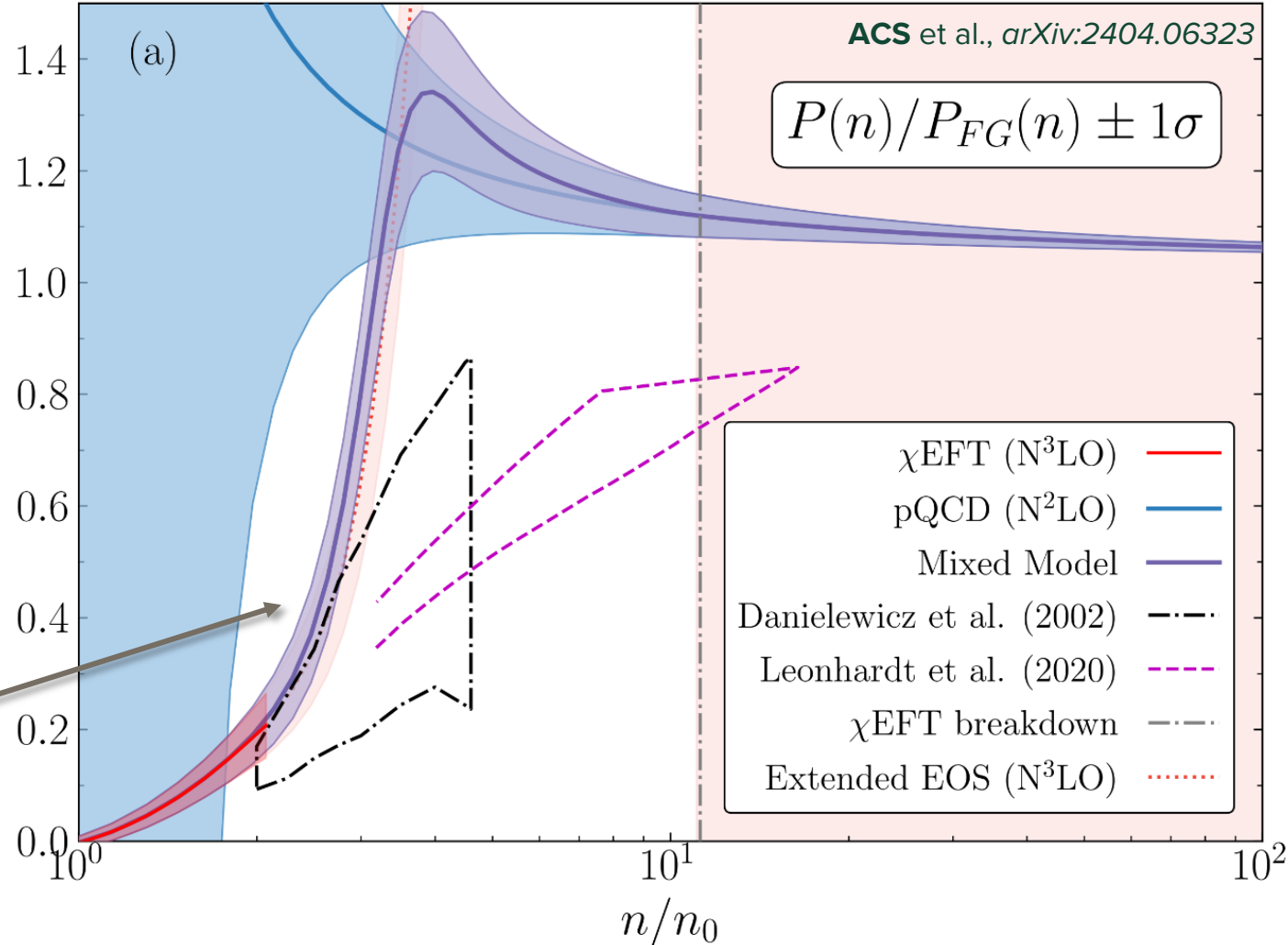
BMM using the pointwise approach

$$\mu = \frac{1}{Z_P} \sum_{i=1}^M \frac{1}{\sigma_i^2} y_i,$$

$$\Sigma^{-1} \equiv \sum_{i=1}^M \frac{1}{\sigma_i^2}$$

Extended EOS from chiral EFT with PAL EOS to have estimates across total density region

Prakash, Ainsworth, Lattimer (1988)



ACS et al., arXiv:2404.06323

$$P(n)/P_{FG}(n) \pm 1\sigma$$



Performed using TOWERET

TOWERET arXiv: 2310.20549

Rapidly increasing EOS from crossover to pQCD

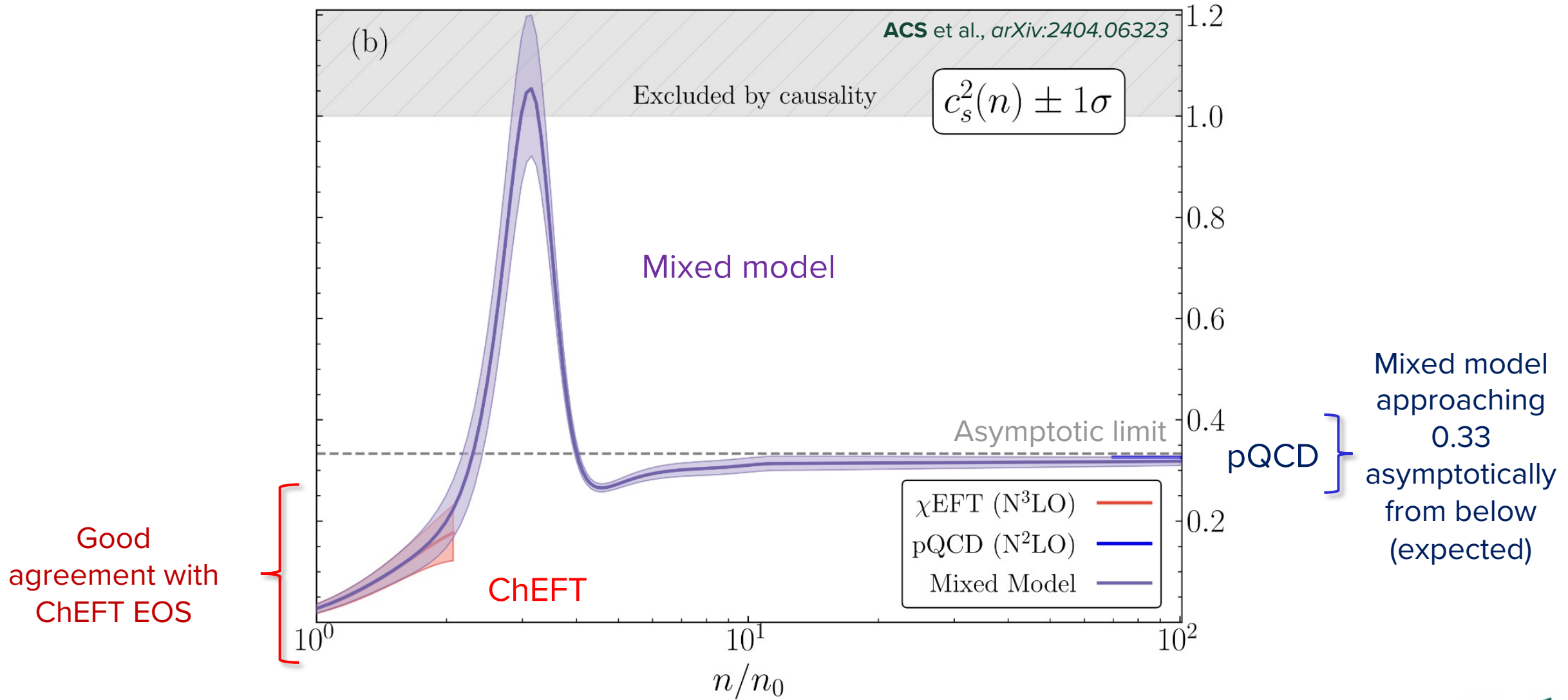
Complete agreement with chiral EFT and pQCD uncertainties in the two limits

Some agreement with heavy-ion data, no overlap with FRG contour

Requires information at all densities from both theories!



# Results: speed of sound



**Acausal EOS**