Constraining the Mass & Radius of Neutron Stars in Globular Clusters

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ABSTRACT
We analyze observations of eight quiescent low-mass X-ray binaries in globular clusters and combine them to determine the neutron star mass-radius curve and the equation of state of dense matter. We determine the effect that several uncertainties may have on our results, including uncertainties in the distance, the atmosphere composition, the neutron star maximum mass, the neutron star mass distribution, the possible presence of a hotspot on the neutron star surface, and the prior choice for the equation of state of dense matter. The distance uncertainty is implemented in a new Gaussian blurring method which can be directly applied to the probability distribution over mass and radius. We find that the radius of a 1.4 solar mass neutron star is most likely from 10 to 14 km and that tighter constraints are only possible with stronger assumptions about the nature of the neutron stars, the systematics of the observations, or the nature of dense matter. Strong phase transitions in the equation of state are preferred, and in this case, the radius is likely smaller than 12 km. However, radii larger than 12 km are preferred if the neutron stars have uneven temperature distributions.

Key words: Dense matter – stars: neutron – X-rays: binaries – Globular clusters

1 INTRODUCTION
With the exception of small corrections from rotation and magnetic fields, the neutron star mass-radius relation is expected to be universal (Lattimer & Prakash 2001). All neutron stars in the universe lie on the same curve, and the determination of that curve informs the study of many neutron star phenomena. The Tolman-Oppenheimer-Volkov equations provide a one-to-one correspondence between the mass-radius relation and the equation of state (EOS) of dense matter, a quantity directly connected to quantum chromodynamics, the theory of strong interactions. In particular, the mass-radius curve is connected to the relationship between pressure and energy density. Since neutron star temperatures are expected to be much smaller than the Fermi momentum of the particles which comprise the neutron star core, neutron stars probe the EOS at zero temperature. There is strong interest in both determining the mass-radius curve from observations and determining the equation of state of cold and dense matter from nuclear experiments and theory.

A critical parameter is the nuclear saturation density, the central density of matter inside atomic nuclei. Nuclear experiments and nuclear theory are extremely successful at determining the nature of matter at and below the saturation density. On the other hand, experiments which probe matter more dense than the saturation density are limited by the fact that they do so at the cost of introducing a large temperature (see, e.g., Tsang et al. 2012). Theory is also currently limited to lower densities: uncertainties are only well-controlled where the Fermi momentum is small enough to employ chiral effective theory or accurate phenomeno-
logical interactions calibrated to nuclei (see recent reviews Gandolfi et al. 2015; Hebeler et al. 2015).

The central density of all neutron stars, however, is likely to be larger than four times the nuclear saturation density (Steiner et al. 2015; Watts et al. 2016). Thus, unless there is a dramatic advance which enables one to construct the cold EOS from experiments which probe hot and dense matter, or there is an unexpected dramatic improvement in nuclear theory-based calculations of dense matter, observations of neutron star masses and radii are likely to be the best probe of cold and dense matter.

1.1 The quiescent LMXB method to constrain NS radii

One promising method to constrain the NS radius is spectral fitting of NSs in low-mass X-ray binaries during periods of little or no accretion, called “quiescence”. Low-mass X-ray binaries (LMXBs) are binary systems containing a NS (or a black hole; we will not discuss those systems here) and a low-mass star (less than 1-2 times the mass of our Sun), where the orbit is tight enough that material can be pulled from the low-mass star down onto the NS. In the majority of LMXBs, the material falling from the companion star piles up in an accretion disk around the NS, where it builds up for months to years until the disk becomes dense and hot enough to become partly ionized, leading to increased viscosity and flow of matter down onto the NS (e.g. Lasota 2001). During these “outbursts”, the falling material converts its potential energy into radiation, primarily in X-rays where the LMXBs typically radiate many 1000s of times the bolometric luminosity of our Sun. These outbursts are detectable across the Galaxy with the use of (low-sensitivity) all-sky X-ray monitors.

Between outbursts, as the disk builds up, the NS is much dimmer, radiating ∼1-100% of the Sun’s bolometric luminosity (10^{31}-10^{33} erg/s). During quiescence, the NS emits heat deposited in the crust and core during outbursts as blackbody-like radiation (Brown et al. 1998). The (ionized) atmosphere quickly stratifies, with the lightest accreted element on top; this topmost layer (typically hydrogen) determines the details (spectrum, angular dependence) of the emitted radiation field (Zavlin et al. 1996; Rajagopal & Romani 1996). In addition to this thermal, blackbody-like radiation, quiescent NS LMXBs often also produce nonthermal X-rays, which can typically be modelled (in the 0.5-10 keV band) with a power-law of photon index 1-2. The nature of the nonthermal X-rays are not clear, though they appear to generally be produced by low-level accretion in quiescence (Campana et al. 1998; Chakrabarty et al. 2014). Continued accretion can also produce thermal blackbody-like emission (Zampieri et al. 1995; Deufel et al. 2001; Rutledge et al. 2002a; Cackett et al. 2010; Bernardini et al. 2013).

By measuring the X-ray flux and temperature of an object at a known distance, the radius of the emitting object can be calculated. Including the redshifting effects of general relativity means that the quantity actually measured is the radius as seen at infinity, $R_\infty = R(1+z) = R/\sqrt{1-2GM/Rc^2}$, such that the outcome is a constrained strip across the mass-radius plane. Since the final spectrum has different dependences on the surface gravity in the atmosphere and on the redshift, it is possible that future, larger effective-area missions may tightly constrain both mass and radius.

Work in this direction has concentrated on quiescent LMXBs in globular clusters for three reasons. First, quiescent LMXB radius measurements depend on knowing the distance, since to first order we constrain the quantity (R/D). Typical LMXBs in our galaxy have poorly known distances (factors of 2 are not uncommon), while globular cluster distances can be known as well as ∼6% (e.g. Woodley et al. 2012). Dense globular clusters produce close accreting binaries in dynamical interactions (e.g. Benaque&a & Downing 2013), making them excellent targets to search for quiescent LMXBs (identifiable through their unusual soft spectra, Rutledge et al. 2002b). Finally, unlike the majority of quiescent LMXBs found outside globular clusters (through recent outbursts), quiescent LMXBs identified in globular clusters tend to have relatively simple spectra, dominated by thermal surface emission with little or no power-law component (Heinke et al. 2003b).

A number of quiescent LMXBs have been studied in some depth with the Chandra and/or XMM-Newton observatories, of which several provide potentially useful constraints on mass and radius. These include quiescent LMXBs in the globular clusters ω Cen (Rutledge et al. 2002b; Webb & Barret 2007; Heinke et al. 2014), NGC 6397 (Grindlay et al. 2001; Guillot et al. 2011; Heinke et al. 2014), M28 (Becker et al. 2003; Servillat et al. 2012), M13 (Gendre et al. 2003; Webb & Barret 2007; Catuneanu et al. 2013), NGC 6304 (Guillot et al. 2009a, 2013), and M30 (Lugger et al. 2007; Guillot & Rutledge 2014). Deep observations of the relatively bright (few 10^{33} erg/s) quiescent LMXB X7 in 47 Tuc gave apparently tight constraints and a large inferred radius (Heinke et al. 2006a), but suffered significantly from an instrumental systematic uncertainty, pileup. Pileup occurs at relatively high count rates, when the energy deposited from two photons is incorrectly recorded as coming from one photon (Davis 2001); some combined photons are interpreted as signals from cosmic rays, and rejected. Although it is possible to model the effects of pileup, this modeling introduces systematic errors that are difficult to quantify; Guillot et al. (2013) pointed out that due to the large fraction (~20%) of piled-up photons in X7 in these data, these systematics were quite large. Recently, new Chandra observations of 47 Tuc X7 in a mode using a shorter frame time to dramatically reduce pileup have provided more reliable radius constraints (Bogdanov et al. 2016).

1.2 Previous combined constraints using quiescent LMXBs

Since each quiescent LMXB provides a constraint covering a large range of mass and radius, several groups have sought to combine constraints from several systems to constrain the locus of mass and radius points for neutron stars. These works have made different assumptions about the composition of the NS atmospheres, and used different methods to combine the constraints from different systems. Guillot et al. (2013) simultaneously fit five quiescent LMXB spectra, forcing all five to have the same radius, arguing for a NS radius of 9.1^{+1.3}_{-1.5} km for the set. The choice of a single radius for all was motivated by the finding that the equation of state curves for most plausible NS structures are nearly vertical
for much of the range of measured NS masses (Lattimer & Prakash 2001). This analysis used a Markov-chain Monte Carlo method to sample the parameter space, and allowed uncertainties in the distances to the globular clusters, variation in the extinction column to each source, and for the possible presence of a hard power-law spectral component (even if not confidently detected).

A key assumption in Guillot et al.’s work was that all quiescent LMXBs have pure hydrogen atmospheres. This is reasonable if the donor stars are hydrogen-rich, since the accreted elements will stratify in less than a minute (Alcock & Illarionov 1980; Hameury et al. 1983), unless accretion continues at such a rate as to replace the photosphere in this time (Rutledge et al. 2002a) Such an accretion rate would produce an accretion-derived X-ray luminosity of order $10^{33}$ erg/s; since accretion is fundamentally a variable process, the lack of detected variability on timescales of years to decades in most globular cluster quiescent LMXBs (including the objects used in these analyses) argues that the thermal emission in these objects comes from stored heat in the NS, and thus that the accretion rate is low enough that the atmosphere is stratified (Heinke et al. 2006a; Walsh et al. 2015; Bahramian et al. 2015).

However, the donor stars may not be hydrogen-rich. Between 28 and 44% of luminous globular cluster LMXBs have orbital periods less than 1 hour (Bahramian et al. 2014), indicating that these NSs accrete from white dwarfs, and suggesting that a similar fraction of quiescent LMXBs may also have white dwarf donors. Detailed calculations of plausible evolutionary scenarios show that the transferred mass will be devoid of hydrogen, and dominated by helium, or carbon and oxygen, depending on the composition of the donor star (Nelemans & Jonker 2010). (Note that binary evolution starting with a slightly evolved secondary star, which would be likely to contain some hydrogen in the transferred mass, cannot explain the observed period distribution of short-period LMXBs in globular clusters; e.g. van der Sluys et al. 2005.) It is possible that accreted matter could spallate nuclei on impact, releasing protons (Bildsten et al. 1993), though spallation might require infalling protons (in’t Zand et al. 2005), and there is no evidence of spallation-produced H in observed thermonuclear bursts in extremely short-period LMXBs (Cumming 2003; Galloway et al. 2008). Finally, diffusive nuclear burning may also consume the hydrogen at the photosphere (Chang & Bildsten 2004).

Since helium and carbon atmospheres shift the emitted X-ray spectra to slightly higher energies with respect to hydrogen atmospheres, the inferred radii (if fit with hydrogen atmospheres) would be smaller than the true radii (Rajagopal & Romani 1996; Ho & Heinke 2009). For carbon atmospheres, the radius difference (a factor of ~two) is large enough that identification should be immediate (none have yet been seen), but the effects of helium atmospheres are more subtle. Several works have considered fits of specific quiescent LMXBs to either H or He atmospheres (Servillat et al. 2012; Catuneanu et al. 2013; Heinke et al. 2014), finding increases of the radius from the fits of ~20-50%.

Other works have combined the individual results for quiescent LMXBs in a Bayesian formalism. Steiner et al. (2010a,b) combined mass-radius constraints from three thermonuclear burst systems (Özel et al. 2009; Güver et al. 2010a,b) with results from three quiescent LMXBs (Heinke et al. 2006b; Webb & Barret 2007). Steiner et al. used a Bayesian framework to combine the results, introducing a parametrized equation of state (incorporating causality constraints, the minimum NS maximum mass, and the low-density nuclear equation of state), and preferred radii (for 1.4 $M_\odot$ NSs) between 11 and 12 km. Lattimer & Steiner (2014b) used a similar Bayesian framework (accounting for the discovery of a 2 $M_\odot$ NS, which increased the minimal NS maximum mass) to reconsider the results of Guillot et al. (2013). Lattimer & Steiner estimated an average bias in the inferred radius of a He-covered NS applied to H atmosphere models of 33%. They took estimates of the probability distribution functions of Guillot et al. (2013), altered them by allowing He atmospheres and alternative choices of interstellar extinction (both alterations performed by analytically estimating the impact), and combined them, finding a preferred radius range of 10.45–12.66 km.

Özel et al. (2016) also combined thermonuclear burst constraints with quiescent LMXB constraints, using an alternative Bayesian formalism that maps the measured masses and radii to the pressures at three fiducial densities. Özel et al. combined individual constraints on six quiescent LMXBs, assuming a hydrogen atmosphere for all systems except the NGC 6397 system, for which they assumed a helium atmosphere, finding a preferred radius between 10.1 and 11.1 km. Finally, Bogdanov et al. (2016) obtained new constraints on the mass and radius of the quiescent LMXBs X7 and X5 in 47 Tuc, and combined these constraints with the results of Özel et al. (2016) to find a preferred radius in the range 9.9–11.2 km.

The goal of this work is to carefully analyze the quiescent LMXB sample, allowing each quiescent LMXB to have either a hydrogen or helium atmosphere, except when independent evidence indicates a particular composition. We combine these measurements in a Bayesian framework, producing results for different assumptions about the NS equation of state, and different assumptions about the quiescent LMXB population.

### 2 DATA ON INDIVIDUAL QUIESCENT LMXBS

We briefly describe the data used to study each quiescent LMXB, and our best estimates of the distance to each globular cluster. We summarize the key information about these sources in Table 1.

#### 2.1 47 Tuc: X7 and X5

47 Tuc is an excellent target due to its relatively short distance, relatively bright quiescent LMXBs, and relatively low $N_H$ (the amount of interstellar gas absorbing soft X-rays). The core is sufficiently dense with X-ray sources that only Chandra observations can fully resolve the sources. The two relatively bright quiescent LMXBs in 47 Tuc, X5 and X7, are bright enough that deep Chandra observations in 2000 and 2002 suffered substantial (~20%) pileup. We therefore obtained 181 kiloseconds of new Chandra observations in 2014/2015, in an observing mode designed to reduce pileup...
to ~1%. Bogdanov et al. (2016) described these observations (plus five shorter observations from 2000 and 2002 in this mode), and the analysis to derive M-R constraints for X7 and X5; we use those M-R constraints in our combined analysis.

X5 shows eclipses with an 8.67 hour period, and dips, stronger at lower energies, due to varying local extinction (Heinke et al. 2003a). These indicate that X5 is viewed nearly edge-on, so that the companion star eclipses X-rays from the NS, and that material from the dynamic accretion disk is often present along the line of sight.

Bogdanov et al. (2016) also compiled a list of literature determinations of the distance to 47 Tuc, by a variety of methods, and compiles a weighted mean of the distance determinations. To limit the effect of flaws in any one method, they conducted “jackknife” tests, where they removed all measurements taken with one method, to assess the systematic errors, finding a final distance of 4.53^{+0.08}_{-0.04} kpc for 47 Tuc.\(^1\) We use Bogdanov et al’s distance estimate, but assume symmetric errors, taking 0.06 kpc as the 1-sigma error uncertainty.

2.2 ω Cen

ω Cen is a relatively nearby and low-density globular cluster, for which either Chandra or XMM-Newton can resolve the known quiescent LMXB. We use Chandra data on ω Cen from 2000 (69 kiloseconds) and 2012 (225 kiloseconds), along with the XMM-Newton data from 2001 (40 kiloseconds), reduced as described by Heinke et al. (2014).

It is often easier to establish the relative distance scale of several globular clusters than an absolute scale. We use the well-established relative difference in distances between ω Cen and 47 Tuc (ω Cen is 16±3% farther than 47 Tuc, Bono et al. 2008) to calculate the distance to ω Cen as 5.22±0.17 kpc. This is consistent with the horizontal branch distance of 5.2 kpc of Harris (1996)\(^2\), the dynamical distance estimate of 5.19^{+0.07}_{-0.08} kpc of Watkins et al. (2015), and in the middle of the other distance estimates discussed by Heinke et al. (2014).

2.3 NGC 6397

NGC 6397 is the second nearest globular cluster, with a very dense core. We use Chandra observations taken in 2000 (49 kiloseconds), 2002 (55 kiloseconds), and 2007 (240 kiloseconds), reduced as described by Heinke et al. (2014) (similar to Guillot et al. 2011)).

As for ω Cen, we use the relative distance measurements to NGC 6397 and 47 Tuc (that NGC 6397 is at 54.5±2.5% of 47 Tuc’s distance, Hansen et al. (2013)) to calculate NGC 6397’s distance at 2.47±0.07 kpc. This is in agreement with the dynamical distance estimate of 2.39^{+0.13}_{-0.11} kpc of Watkins et al. (2015), and with most other recent distance estimates discussed in Heinke et al. (2014).

2.4 M13

M13 is a relatively low-density cluster with very low extinction. It was the subject of extensive ROSAT observations (46 kiloseconds) in 1992 with the PSPC camera, which accurately measured the absorption of the low-energy spectrum (though ROSAT has much poorer spectral resolution in general). We use this ROSAT PSPC spectrum, 2002 XMM-Newton observations (34.6 kiloseconds), and 2006 Chandra observations (54.7 kiloseconds), reduced as described by Catuneanu et al. (2013) (similar to Webb & Barret (2007)).

The distance to M13 has been extensively discussed by Sandquist et al. (2010), who estimate 7.65±0.36 kpc (reconciling distance estimates using the tip of the red giant branch with horizontal branch estimates), while the homogeneous relative distance estimates of Recio-Blanco et al. (2005) (which perfectly matches our distance estimate to 47 Tuc) gives a distance of 7.8±0.1 kpc. We use Recio-Blanco’s distance, but conservatively use the larger distance uncertainty, 7.8±0.36 kpc.

2.5 NGC 6304

The quiescent LMXB in NGC 6304 is located in the core of its cluster near other sources (Guillot et al. 2009b), so we use only the deep Chandra observation of 2010 (98.7 kiloseconds), extracted following Guillot et al. (2013), using CIAO 4.7 and CALDB 4.6.9.

The distance to NGC 6304 is uncertain due to its relatively high extinction, and thus high uncertainty on its extinction. Using the distance modulus of Recio-Blanco et al. (2005) and the extinction estimate of Piotti et al. (2002) (following Guillot et al. 2013) gives a distance of 6.22±0.26 kpc.

2.6 M30

The quiescent LMXB lies in the extremely dense core of the distant, but low-extinction, globular cluster M30, and can only be resolved by Chandra. We use the 2001 Chandra observation (49 kiloseconds), and extract the data following

<table>
<thead>
<tr>
<th>Cluster</th>
<th>Distance (kpc)</th>
<th>(N_H) (cm(^{-2}))</th>
<th>References</th>
</tr>
</thead>
<tbody>
<tr>
<td>47 Tuc</td>
<td>4.53±0.06</td>
<td>3.5 \times 10^{20}</td>
<td>1,2,3,4</td>
</tr>
<tr>
<td>ω Cen</td>
<td>5.22±0.17</td>
<td>1.3 \times 10^{21}</td>
<td>5,6,4</td>
</tr>
<tr>
<td>NGC 6397</td>
<td>2.47±0.07</td>
<td>1.2 \times 10^{21}</td>
<td>3,4,5</td>
</tr>
<tr>
<td>M13</td>
<td>7.8±0.36</td>
<td>1.1 \times 10^{20}</td>
<td>7,8,4</td>
</tr>
<tr>
<td>NGC 6304</td>
<td>6.22±0.26</td>
<td>4.6 \times 10^{21}</td>
<td>7,9,10</td>
</tr>
<tr>
<td>M30</td>
<td>8.78±0.33</td>
<td>2.6 \times 10^{20}</td>
<td>7,11,4</td>
</tr>
<tr>
<td>M28</td>
<td>5.5±0.3</td>
<td>2.3 \times 10^{21}</td>
<td>12</td>
</tr>
</tbody>
</table>

Lugger et al. (2007) (also Guillot & Rutledge 2014), using CIAO 4.7 and CALDB 4.6.9.

Recio-Blanco et al. (2005) measure a distance of 8.78±0.33 kpc; since their distance estimate aligns with ours for 47 Tuc, we adopt this. We note that it is also in agreement with other estimates by, e.g., Dotter et al. (2010) (8.8 kpc).

2.7 M28

The quiescent LMXB in M28 (source 26 of Becker et al. 2003) lies in the core of this relatively nearby, dense cluster. We use three Chandra observations from 2002 (42 kiloseconds) and two from 2008 (199.6 kiloseconds), reduced as described in Servillat et al. (2012), using CIAO 4.7 and CALDB 4.6.9.

We use the estimate of the distance to M28 of 5.5±0.3 kpc, derived from the brightness of the horizontal branch (Testa et al. 2001) calibrated by Harris (1996, 2010 revision).

3 X-RAY SPECTRAL FITTING

For the spectral fitting, we used the XSPEC software (Arnaud 1996). We experimented with merging spectra taken with the same detector close in time, or fitting them simultaneously; with grouping the data into bins of >20, >50, or other numbers of counts; with analyses using χ² or C-statistics. In general, such changes do not make substantial differences to the final results.

For all spectra, our spectral fits included a neutron star atmosphere, either NSATOMS (hydrogen, Heinke et al. 2006a) or NSX (helium, Ho & Heinke 2009). We included $N_H$ through the TBABS model (with the extinction free to vary), using element abundances from Wilms et al. (2000) and photoelectric cross-sections from Verner et al. (1996).

We include a power-law in the spectral fitting, although it is not required for any quiescent LMXB, with the photon index fixed to 1.5, as typical for power-law components in quiescent LMXBs (Campana et al. 1998; Cackett et al. 2010; Chakrabarty et al. 2014). We add systematic errors, of magnitude 3%, to all spectra, accounting for instrumental calibration uncertainties, following Guillot et al. (2013); Bogdanov et al. (2016). Our spectral fits were performed assuming the nominal best-fit distances, with distance uncertainties convolved with the probability density functions (from the spectral fits) during the Bayesian MCMC calculation (see below).

Bogdanov et al. (2016) found that the inclusion of pileup in spectral modelling for Chandra observations of X7 in 47 Tuc made a significant difference to the final radius contours, even at pileup fractions as low as 1%. For this reason, we include pileup in all Chandra spectral fits; this is particularly relevant for the NGC 6397 spectral fits, since previous fits (Guillot et al. 2013; Heinke et al. 2014) did not include pileup for this source. The M28 quiescent LMXB has the highest fraction of piled-up events, with about 5% of photons piled up (Servillat et al. 2012).

A crucial uncertainty is the chemical composition of the atmosphere of the quiescent LMXB. In the Bayesian MCMC analysis below, we analyze the spectra for each quiescent LMXB with both hydrogen and helium atmospheres with two exceptions. The first exception is the quiescent LMXB ω Cen which has a firm detection of hydrogen in its spectrum (Haggard et al. 2004). The second is X5, which has a long orbital period (Heinke et al. 2003a) suggesting a hydrogen-rich donor (see section 1.2).

3.1 Systematic Uncertainties

There are remaining systematic uncertainties which we have not robustly controlled. Perhaps the largest uncertainty is the possible effect of hot spots upon the inferred radius. Many neutron stars show pulsations, implying the presence of hotter regions on their surface; examples include young pulsars (De Luca et al. 2005), old millisecond pulsars (Bogdanov 2013), young neutron stars without pulsar activity (Gottghelf et al. 2010), and accreting neutron stars (Patruno et al. 2009). Hot spots may be produced by the accretion of material onto a magnetic pole, collision of relativistic electrons and positrons with the pole during pulsar activity, or preferential leakage of heat from the core along paths with particular magnetic field orientations (Potekhin & Yakovlev 2001). While these mechanisms depend on the presence of a magnetic field, the field strengths in quiescent LMXBs and the details of these mechanisms are not known well enough to fully rule out the possibility of a hot spot. Elshamouty et al. (2016) explored the effect of hot spots on quiescent LMXB spectra, focusing on the cases of X7 and X5 in 47 Tuc. Elshamouty et al. used deep Chandra observations at high time resolution to search for pulsations from these two sources. Considering a range of possible magnetic latitudes and inclinations, they derived limits on the hot spot temperature, and then derived limits on the bias that might be inferred by fitting a single-temperature neutron star atmosphere to a neutron star that has hot spots. Unfortunately, existing data poses only weak limits, such that the spectroscopically inferred radius could be biased downwards up to 28% smaller than the true radius. We will consider biases up to this level in some of our analyses.

Another concern is variability in the absorbing column. As mentioned above, X5 in 47 Tuc is a nearly edge-on system that shows varying photoelectric absorption. Bogdanov et al. (2016) shows that if periods of enhanced absorption (signified by dips in the count rate) are removed, the inferred radius of X5 grows significantly. When time periods including different absorption values are combined and fitted with a single absorber, the spectrum appears intrinsically more curved (and thus at a hotter temperature). However, we cannot be sure that we have removed all the periods of enhanced photoelectric absorption; short periods of enhanced absorption would not supply enough counts to enable unambiguous determination of a dip. We thus suspect that the spectral fits to X5 may be biased downwards by varying photoelectric absorption. This problem might affect other quiescent LMXBs as well. For typical mass ratios of ~0.3, the fraction of LMXBs that should show eclipses (these systems tend to show dips) is around 20-25%. However, the neutron stars in the most edge-on systems will be continually obscured by the accretion disk, so the fraction of detectable quiescent LMXBs that can show eclipses and dips should be roughly 10%. Among our eight targets, one (X5) indeed shows both these behaviours; the lack of detectable eclipses in other systems suggests that dips are not likely. However, absorption
dips have been seen in some LMXBs that are not edge-on (cf. Galloway et al. 2016; Mata Sánchez et al. 2016), although to date these dips have only been seen during outburst. It is not clear if substantial absorption off-plane is likely when the systems are in their quiescent state.

4 BAYESIAN INFERENCE FOR NEUTRON STAR MASSES AND RADI

Ideally, one directly connects the probability distributions of the quantities of interest to the observables. In this case, that would imply directly connecting neutron star masses and radii to the flux of photons at every energy. In practice, however, this would require an integral over several energy bins for each neutron star data set. In this work, we convert the X-ray spectrum for each source into a probability distribution for the neutron star with radius $R$, mass $M$, at distance $D$ with atmospheric composition $X$.

\[ P(Q) \propto \int \left\{ \prod_{i=1}^{M} D_i[R(M_i, \{p_i\}), M_i, D_i, X_i] \right\} \times \delta[q - Q(p_1, \ldots, p_N, M_1, \ldots, M_N, D_1, \ldots, D_N, X_1, \ldots, X_N)] \times dp_1 \ldots dp_N dM_1 \ldots dM_N \times dD_1 \ldots dD_N dX_1 \ldots dX_N, \]  

(2)

where $N$ is the number of neutron stars being considered. Several examples of relevant posteriors are reviewed briefly in Lattimer & Steiner (2014a).

We assume the composition, $X_i$, of the atmosphere of each neutron star (except for the neutron star in $\omega$ Cen and for X5 in 47 Tuc) is a discrete binary variable: either H or He. As a prior distribution we assume a 2/3 probability of H and a 1/3 probability of He, following the observed ratio of H-rich to He-rich donors in bright LMXBs in globular clusters (Bahramian et al. 2014, and references therein). This method is superior to that employed in Lattimer & Steiner (2014b) because it allows us to quantitatively predict the posterior probability that any particular neutron star has a H or He atmosphere. On the other hand, it has been argued by some authors that helium atmospheres are expected to be unlikely (Guillot & Rutledge 2014), so we also try models where the prior probability for hydrogen is 90% or 100%.

We remove the neutron star in X5 from our baseline data set because of the varying absorption described in section 3.1. We assume the maximum mass must be larger than $2 M_\odot$ as implied by the recent observations of high mass neutron stars (Demorest et al. 2010; Antoniadis et al. 2013). We allow neutron stars to have masses as low as 0.8 $M_\odot$ (which may be conservative), but increasing this number has little effect on our results.

As discussed in Bogdanov et al. (2016) (see also Besnoygov & Yakovlev 2015), there are strong reasons to believe that these neutron stars are not close to 2 $M_\odot$ in mass, since that would likely produce extremely rapid cooling (by processes such as direct Urca, e.g. Yakovlev et al. 2003) so that we would not observe strong thermal radiation from their surfaces. This method is potentially powerful to constrain the masses of neutron stars, but at this time the mass at which rapid cooling turns on is not well-constrained. The neutron star in SAX J1808.4-3658 is extremely cold, indicating that rapid cooling processes are active (e.g. Heinke et al. 2009). This neutron star seems not to have a particularly large mass (the 2$\sigma$ upper limit on the mass is $\sim 1.7 M_\odot$), though current constraints are not very strong (Wang et al. 2013).

4.1 Distance and Hotspot Correction

In this work, we choose to marginalize over the distance as a nuisance variable instead of producing a separate fit for each distance. So long as all of the probability distributions of interest (all of the quantities $P_2$ in eq. 2 above) are independent of distance, we can perform the distance integrations first. This means that we cannot generate any posterior distributions for the distance, but we expect other methods to provide superior distance measurements anyway.

There is a bijection between $(R, M)$ and $(R_\infty, z)$ when $R > 2GM$ (we take $c = 1$ and suppress the composition dependence for now):

\[ R_\infty(R, M) = R \left(1 - \frac{2GM}{R}\right)^{-1/2} \]  

(3)

and

\[ z = \left(1 - \frac{2GM}{R}\right)^{-1/2} - 1. \]  

(4)

The opposite transformation is given by

\[ M(R_\infty, z) = \frac{R_\infty}{c} \left[\frac{z(2 + z)}{2(1 + z)^3}\right] \]  

(5)

and

\[ R(R_\infty, z) = \frac{R_\infty}{(1 + z)}. \]  

(6)
To first order (adequate for small fractional uncertainties), distance scales with $R_\infty$, thus $R_{\infty,new}/D_{\text{new}} = R_{\infty}/D_{\text{old}}$ and the fractional uncertainty in $R_\infty$ is $\delta R_\infty = (\delta D/D_{\text{old}})R_\infty$. To correct for a variation in distance, $D = D_{\text{new}} \pm \delta D$ kpc given the original distance $D_{\text{old}}$, we apply a Gaussian blurring of the value of $R_\infty$ which can be represented by the replacement of the initial two-dimensional probability distribution
\[
P(R_\infty, M) \rightarrow \int C d\tilde{R}_\infty P(\tilde{R}_\infty, M) \exp \left[ -\frac{(\tilde{R}_\infty - R_{\infty,new})^2}{2(\delta R_\infty)^2} \right]
\]
where $C$ is a normalization constant and $\tilde{R}_\infty$ is a dummy integration variable. Thus, in order to correct for the distance uncertainty, we transform from $\mathcal{D}(R, M)$ to $\mathcal{D}(R_\infty, z)$, perform the Gaussian blurring from Eq. 7, and then transform back to $\mathcal{D}_{\text{new}}(R, M)$. Using the definitions above and replacing $R_{\infty,new}$ and $\delta R_\infty = R_\infty \delta D/D_{\text{old}}$ with their expressions from above, we obtain
\[
\mathcal{D}_{\text{new}}(\hat{R}, \hat{M}) = \int_0^{\infty} d\tilde{R}_\infty \left[ \frac{D_{\text{old}}}{R_\infty(\hat{R}, \hat{M}) \delta D \sqrt{2\pi}} \right]
\exp \left\{ -\frac{[\tilde{R}_\infty - R_{\infty} (\hat{R}, \hat{M}) D_{\text{new}}/D_{\text{old}}]^2}{2[D_{\text{new}}/D_{\text{old}}]^2} \right\} \times D_{\text{old}}[R[R_{\infty}, z(\hat{R}, \hat{M})], M[\tilde{R}_\infty, z(\hat{R}, \hat{M})]],
\]
where a caret is used to distinguish the values $\hat{R}$ and $\hat{M}$ from the functions in Eqs. 5 and 6 above. To see that this expression makes sense, take the limit $\delta D \to 0$, then the exponential and associated normalization is equal to a delta function,
\[
\left[ \frac{D_{\text{old}}}{R_\infty(\hat{R}, \hat{M}) \delta D \sqrt{2\pi}} \right] \times \exp \left\{ -\frac{[\tilde{R}_\infty - R_{\infty} (\hat{R}, \hat{M}) D_{\text{new}}/D_{\text{old}}]^2}{2[D_{\text{new}}/D_{\text{old}}]^2} \right\} \delta \rightarrow 0
\]
so in this limit
\[
\mathcal{D}_{\text{new}}(\hat{R}, \hat{M}) = D_{\text{old}}[R[R_{\infty}, (\hat{R}, \hat{M}) D_{\text{new}}/D_{\text{old}}, z(\hat{R}, \hat{M})], M[R_{\infty}, (\hat{R}, \hat{M}) D_{\text{new}}/D_{\text{old}}, z(\hat{R}, \hat{M})]]
\]
If $D_{\text{new}} = D_{\text{old}}$, the term on the right-hand side of Eq. 10 just becomes $D_{\text{old}}(\hat{R}, \hat{M})$ (i.e. the probability distribution is unchanged). Finally, Eq. 8 also ensures that the new relative uncertainty in $R_\infty$ is equal to $\delta D/D_{\text{new}}$, since
\[
\frac{\delta \tilde{R}_\infty}{\tilde{R}_\infty} \rightarrow \frac{R_{\infty}(\hat{R}, \hat{M}) \delta D/D_{\text{old}}}{R_{\infty}(\hat{R}, \hat{M}) D_{\text{new}}/D_{\text{old}}} = \frac{\delta D}{D_{\text{new}}}
\]
Fig. 1 gives a demonstration of the method. The distribution is transformed to $(R_\infty, z)$ space (upper-right panel) where the distance uncertainty is applied (lower-left panel), and then transformed back to $(R, M)$ space (lower-right panel).

Alternatively, we can rescale $R$ at constant $z$ and obtain the same distribution $\mathcal{D}_{\text{new}}$. To show this, create a new dummy integration variable, $\mathcal{R} \equiv R_{\infty}/[1 + z(\hat{R}, \hat{M})]$. The value $z(\hat{R}, \hat{M})$ is independent of the integration variable, $\mathcal{R}$, so $d\mathcal{R}/dR_\infty = 1/[1 + z(\hat{R}, \hat{M})]$. Since $R_{\infty}(\hat{R}, \hat{M})/[1 + z(\hat{R}, \hat{M})] = \mathcal{R}$, the integral on the right-hand side of Eq. 8 can be rewritten to give
\[
\mathcal{D}_{\text{new}}(\hat{R}, \hat{M}) = \int_0^{\infty} d\mathcal{R} \left[ \frac{1}{\mathcal{R} \delta \mathcal{R} \sqrt{2\pi}} \right] \exp \left\{ -\frac{[\mathcal{R} - \hat{R} D_{\text{new}}/D_{\text{old}}]^2}{2[D_{\text{new}}/D_{\text{old}}]^2} \right\} D_{\text{old}}[\mathcal{R}, M[R_{\infty}, z(\hat{R}, \hat{M})]].
\]

Note that, in the figures below, the rescaled results from either Eqs. 8 or 12 shows that the probability distribution vanishes at the extremes in $M$ and $R$. This is because we must assume the input probability distributions from the X-ray fits are step functions (e.g. they drop immediately to zero probability for $R < 9$ km) and these step functions are softened by the additional distance uncertainty.

A demonstration of the method implied by Eq. 12, applied to the neutron star in NGC 6304 is given in the upper panels of Fig. 2. The final result (not shown) is indistinguishable from the lower right panel in Fig. 1. The final results, given by Eq. 8, for the baseline data set and for X5 in 47 Tuc assuming a H atmosphere are presented in Fig. 3. The effect of an He atmosphere on the neutron stars (except for X5 and the neutron star in $\omega$ Cen) is shown in Fig. 4.

Elshamouty et al. (2016) found that the presence of temperature inhomogeneities on the neutron star surface...
4.2 Equation of State

Quantum Monte Carlo calculations of neutron matter (Gandolfi et al. 2012) provide an excellent description of matter up to the nuclear saturation density ($\rho \approx 2.8 \times 10^{14} \, \text{g/cm}^3$). As in Steiner et al. (2015), we assume that neutron star matter near the saturation density is described by results from quantum Monte Carlo. We also assume the neutron star has a crust as described in Baym et al. (1971) and Negele & Vautherin (1973).

High-density matter may contain a strong phase transition. For the purposes of this work, a strong phase transition may be defined as a region of energy density for which the pressure is nearly flat. If matter is unlikely to have a strong phase transition, then a polytropic form is a good description of high-density matter. We employ Model A as presented in Steiner et al. (2013) and updated in Steiner et al. (2015). On the other hand, strong phase transitions are not well-described by polytropes. In pressure-energy density space a polytrope of the form $P = K e^\gamma$ is not a good description of a nearly flat EOS because it requires $\Gamma$ to be very small and $K$ to be anomalously large. This kind of behavior is disfavored by uniform or weakly-varying priors in $\Gamma$. We thus also use Model C (Steiner et al. 2013, 2015) which uses line-segments in pressure and energy density space and makes stronger phase transitions more likely. It is important to emphasize that the ability of the parameterization to describe a generic EOS is not the only consideration (as both polytropic models and those based on line segments can reproduce almost all EOSs). Once a parameterization and a prior for the parameters is specified, changes in the likelihood with which various EOSs are selected can make...
which constrain the neutron star mass more than the base-der to properly evaluate the Bayes factor for the models rescaling; it is important that they are not rescaled in or-the range \( [0, 1] \). The neutron star mass parameters need no

4.3 Model Comparison Using Bayes Factors

In order to compare models, we employ Bayes factors, defined as the ratio of the evidence. The evidence is the integral, over the full parameter space, of the posterior distribution. The normalization of the evidence here requires some care. When the two models which are being compared have the same dimensionality and their parameters have the same units, the Bayes factor needs no extra normalization. However, this is not the case here as Model C and our baseline model have fundamentally different parameters, so the evidence is computed by rescaling the distinct model parameters which have dimensionful units so that they are in the range \([0, 1]\). The neutron star mass parameters need no rescaling; it is important that they are not rescaled in order to properly evaluate the Bayes factor for the models which constrain the neutron star mass more than the base-

line model. Bayes factors between 1/3 and 3 are generally regarded as relatively weak, and in this case no definitive statement can be made about the two models.

5 RESULTS

Ideally, one decides the prior distribution before any calculations are performed. We assign all the different models and interpretations of the data (described in detail below) equal probability. Thus our full posteriors can be viewed as the sum over the results presented for each individual figure.
model and/or data interpretation (after having been properly weighted by their evidence).

For our baseline results, we include all neutron stars except X5, and assume the polytropic model for high-density matter. We include both H and He atmospheres for the objects in the baseline data set, except ω Cen, where the atmosphere composition is known to be H (Haggard et al. 2004). We initially assume the abundances of elements in the interstellar matter to be those given by Wilms et al. (2000), since that paper collated the best available evidence on the abundance of elements in the local interstellar medium. In order to test the change in the inferred radius due to different abundance models, we compare our mass and radius distributions derived using the Wilms et al. (2000) abundance model to those derived using the abundance models of Asplund et al. (2009) and Lodders (2003). These abundance models, produced using studies of the Sun and meteorites, respectively, suggest a plausible range of uncertainty for the interstellar abundances. This is in contrast to significantly older abundance models such as those of Anders & Grevesse (1989), which are quite different, and would lead to significant changes to radius estimates (Heinke et al. 2014). Fig. 7 shows the mass and radius constraints for X7 in 47 Tuc using these alternate abundance models and demonstrates that the results are only slightly different (differences in inferred radius are less than 1%, in agreement with Bogdanov et al. 2016 from those using Wilms et al. (2000) abundances. This holds for all of the objects in our data set, so we only present results using Wilms et al. (2000) abundances below. We have also tested the effects of the updated tbnew code by Jörn Wilms3, and find that the code improvements of tbnew (mostly fine structure around edges, affecting high-resolution spectroscopy) affect radius estimates on the order of 0.1%.

We describe all nine separate combinations of data sets and model assumptions below, highlighting important features for each:

(1) “baseline”: The posterior mass and radius distributions for the neutron stars in the baseline scenario are given in Fig. 8, together with the results presuming Model C is used for the EOS. The mass posterior distributions are relatively broad, with the sole exception for X7. The mass of X7 must be larger than 1.1 M⊙ in Model C because smaller mass stars have central densities too low to allow for the phase transition to decrease the radius sufficiently to match the mass and radius implied by the observations.

The upper left panel of Fig. 9 shows an ensemble of one-dimensional radius histograms for a fixed mass. The results imply that the radius for a M = 1.4 M⊙ neutron star is between 11.0 and 14.3 km (to 95% confidence; see Table 2). The figure gets less dark at higher mass because the area under a radius histogram at fixed mass is normalized to the probability that the maximum mass is larger. Note that the probabilities shift towards larger radii for masses above 2 M⊙ because larger mass stars require a larger maximum mass, which in turn requires a larger pressure at lower densities and thus a larger radius. The probabilities for a helium atmosphere for each source are given in Table 3 and strongly prefer a He atmosphere for the neutron star in NGC 6397 and a H atmosphere for the neutron star X7 in 47 Tuc. Note that the probability of a He atmosphere hovers around 1/3 for objects like the neutron star in NGC 6397, the neutron star in M13, and X7 in 47 Tuc. The probabilities for helium atmospheres drop significantly for some objects, including the neutron star in NGC 6397, the neutron star in M13, and X7 in 47 Tuc.

(2) “Model C”: A significant decrease in the radius comes from using a model of high-density matter which allows for strong phase transitions, as found in Steiner et al. (2013, 2015). This causes likely radii for M = 1.4 M⊙ neutron stars to drop by 1 to 2 km (see also upper right panel of Fig. 9). The probabilities for helium atmospheres drop significantly for some objects, including the neutron star in NGC 6397, the neutron star in M13, and X7 in 47 Tuc.

(3) “H atm.”: Assuming all neutron stars must have a hydrogen atmosphere (lower left panel of Fig. 9) also decreases the 95% confidence limits in the radius by 0.2 (lower bound) and 1.3 km (upper bound).

(4) “Mmax > 2.3”: Requiring the neutron star maximum mass to lie above 2.3 M⊙ increases the lower limit for the radius by 0.7 km (lower right panel of Fig. 9), similar to the result found in Steiner et al. (2013, 2016).

(5 and 6) “1.3 < M < 1.5” and “1.3 < M < 1.7”:

3 http://pulsar.sternwarte.uni-erlangen.de/wilms/research/tbabs/
Constraining the mass of the neutron stars (to either $1.3 \sim 1.7 \, M_\odot$ or to $1.3 \sim 1.5 \, M_\odot$) also has relatively little effect on the inferred radii as shown in the upper left and upper right panels of Fig. 10.

(7) “with X5”: If we include X5 in our data set (lower left panel of Fig. 10), then radii above $13.9$ km for a $M = 1.4 \, M_\odot$ neutron star are strongly ruled out. The lower limit for the radius decreases, but only slightly, as radii smaller than about 11 km require a strong phase transition which are disfavored in polytropic models. In this case, the preference for a helium atmosphere decreases in objects for which the probabilities are not dominated by the prior choice (which tends to be those stars which have posterior probabilities not near $33\%$).

(8) “hotspot”: The model permitting hotspots significantly increases the $1\sigma$ range for the radius (lower right panel of Fig. 10), but has little effect on the $2\sigma$ limits (Table 2). The effect of the hotspot on the posterior probability for the atmosphere is most dramatic for the neutron star in NGC 6397 (see table 3). In this case, the hotspot effectively increases the inferred radius, and this makes the H atmosphere more probable, thus decreasing the posterior probability for a He atmosphere by almost a factor of two.

(9) “90% H”: In the baseline model, we assume a $2/3$ prior probability that each neutron star has a hydrogen atmosphere. Increasing this prior to $90\%$ decreases the posterior probability as seen in the last column of table 3, and the effect of this prior choice on the posterior probability is stronger than our other model choices. The effect on the neutron star radius is more modest, as shown in the last row of Table 2.

The evidence integrals are presented in Table 4, and the Bayes factor for one model with respect to another can be computed by forming the corresponding ratio of the evidence. Requiring all neutron stars to have a H atmosphere is ruled out with a Bayes factor of $1/3$, but again as described above this result will strongly depend on the prior probability of a H atmosphere. Constraining the neutron star mass to between $1.3$ and $1.7 \, M_\odot$ drops the evidence by a factor of 2 and further constraining it to between $1.3$ and $1.5 \, M_\odot$ drops the evidence by an additional $30\%$. Assuming the presence of hotspots drops the evidence by a factor of two. In agreement with previous work, we find that the choice of EOS model has a strong impact (see e.g. Steiner et al. (2015)). We find that the Model C is strongly preferred over the baseline result (a Bayes factor of 8.4). One way of diagnosing which object contributes most strongly to this improved fit is by looking at the ratio of the average posterior probability for each object between Model C and the baseline result. These ratios are $1.03, 2.24, 1.03, 1.39, 0.94, 1.29,$ and $1.86$ for the neutron stars in NGC 6304, NGC 6397, M13, M28, M30, w Cen, and for X7 in 47 Tuc, respectively. (The product of these ratios is not exactly equal to 8.4 because of correlations between the weight from each neutron star). Thus the neutron star in NGC 6397 most strongly
Table 2. Two-sigma confidence limits for the radius of a 1.4 $M_\odot$ neutron star (in km) for the different combinations of data sets and model assumptions used in this work.

<table>
<thead>
<tr>
<th>Source</th>
<th>Probability of He atmosphere</th>
<th>Evidence</th>
</tr>
</thead>
<tbody>
<tr>
<td>NGC 6304</td>
<td>32 % 32 % 30 % 31 % 9.1%</td>
<td>2.4</td>
</tr>
<tr>
<td>NGC 6397</td>
<td>81 % 70 % 47 % 41 % 34 %</td>
<td>0.68</td>
</tr>
<tr>
<td>M13</td>
<td>28 % 26 % 19 % 26 % 7.0%</td>
<td>0.88</td>
</tr>
<tr>
<td>M28</td>
<td>52 % 48 % 31 % 33 % 16 %</td>
<td>1.31</td>
</tr>
<tr>
<td>M30</td>
<td>31 % 30 % 27 % 31 % 8.7%</td>
<td>1.0</td>
</tr>
<tr>
<td>X7</td>
<td>16 % 2.5 % 3.4 % 13 % 2.0%</td>
<td>2.5</td>
</tr>
</tbody>
</table>

Table 3. Probability of a Helium atmosphere for each neutron star depending on data set and model assumptions. The statistical uncertainties in these probabilities are about 3%.

<table>
<thead>
<tr>
<th>Data and model selection</th>
<th>Evidence</th>
</tr>
</thead>
<tbody>
<tr>
<td>baseline</td>
<td>2.4</td>
</tr>
<tr>
<td>w/X5</td>
<td>0.68</td>
</tr>
<tr>
<td>Model C hotspot</td>
<td>0.88</td>
</tr>
<tr>
<td>90% H</td>
<td>1.31</td>
</tr>
</tbody>
</table>

Table 4. The evidence, as computed by the properly normalized integral under the posterior distribution, for the different combinations of data sets and model assumptions used in this work.

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rather low densities to match the stars which have smaller inferred values of $R\infty$ and the pressure increases strongly above the phase transition in order to ensure that the maximum mass is above 2 $M\odot$. The lower left panel assumes H atmospheres, leading to a smaller pressure in the range of energy densities from 300 to 500 MeV/fm$^3$. Increasing the maximum mass, shown in the lower right panel, has the opposite effect. In the lower right panel, the central energy density of the maximum mass star decreases significantly, as one expects when increasing the maximum mass. Note that, since the posterior mass for the neutron stars in our data set is not often near 2 $M\odot$, our constraint on the EOS at high densities will be more dependent on the prior distribution and the parameterization of the EOS. The lower and upper limits for the pressure over a range of energy densities for all nine scenarios is given in Table 5.

![Figure 11. A set of posterior distributions for the pressure at fixed energy density over a range of energy densities for some of the models used in this work. The upper left panel shows results for our baseline model, the upper right shows the results for Model C, the lower left assumes H atmospheres for all sources and the lower right assumes a maximum mass larger than 2.3 $M\odot$.](image)

The neutron star central baryon density is an important parameter for understanding the extent to which nuclear physics can play a role in neutron star structure. We find that the central baryon density of the maximum mass star could be as low as 4 times the nuclear saturation density, $n_0 = 0.16$ fm$^{-3}$ or as large as $8n_0$, in agreement with Steiner et al. (2015). The masses and radii for neutron stars which have central baryon densities equal to integer multiples of $n_0$ are shown in Fig. 12. The left panel shows our results for the baseline model and the right panel shows the results for Model C. This figure is an updated version of Fig. 2 from Gandolfi et al. (2012) which properly quantifies the uncertainty in the result by removing the unwarranted assumption that matter at high densities can be described by the same Hamiltonian which is used at the nuclear saturation density. The relationship with the density derivative of the symmetry energy, $L$, is also shown. It is standard to write the energy density $\varepsilon(n_B, \delta)$ as a function of the neutron and proton number densities, $n_n$ and $n_p$, using the definitions $n_B = n_n + n_p$ and $\delta = 1 - 2n_p/n_B$. Then the quantity $L$ is defined by

$$L \equiv 3n_B \frac{\partial}{\partial n_B} \left( \frac{1}{2n_B} \frac{\partial^2 \varepsilon}{\partial^2 \delta} \right).$$

Lattimer & Prakash (2001) showed that for a set of typical models $L$ is correlated with the radius of a 1.4 solar mass neutron star. However, this correlation is weak if one assumes a prior probability which allows for strong phase transitions in the EOS (Steiner et al. 2016) as also shown in Fig. 12. The lower and upper limits for the value of $L$...
and for the radii of higher mass stars for all nine scenarios is given in Table 6.

The one- and two-sigma limits for the radius of the maximum mass neutron star are given in Table 7. The most extreme radii are observed if the EOS at high densities is assumed to have strong phase transitions, in which case the two-sigma limit goes below 9.4 km. Larger lower limits occur if the maximum mass is greater than 2.3 $M_\odot$ or if the neutron stars have uneven temperature distributions, otherwise the lower two-sigma limit is typically around 9.7 km. Table 7 also presents the limits for the central baryon (ranging from 0.61 to 1.3 fm$^{-3}$) and energy density (ranging from 740 to 1700 MeV/fm$^3$). Increasing the maximum mass tends to decrease these central densities significantly.

### 6 DISCUSSION

Steiner et al. (2013) found radii between 10.4 and 12.9 km (to 95% confidence) for a 1.4 $M_\odot$ neutron star, and our results allow for a larger range of radii. This is principally because, except for the neutron star in $\omega$ Cen and X5 in 47 Tuc, we have relaxed the assumption that quiescent LMXB atmospheres are composed of hydrogen. Once we assume quiescent LMXBs have hydrogen atmospheres, we reproduce the previous result. Our model with the largest evidence, Model C, suggests that the radius of a 1.4 $M_\odot$ neutron star is less than 12 km to 95% confidence. One could conceive of many possible combinations among the model assumptions which we have explored. For example, if the maximum mass were larger than 2 $M_\odot$, and additionally quiescent LMXBs all had uneven temperature distributions, then their radii could be larger than 14 km, especially if strong phase transitions were ruled out by theoretical work on the nucleon-nucleon interaction. Alternatively, we would find even smaller radii than 12 km if we assumed that He atmospheres are unlikely and the data from X5 was confirmed (a scenario similar to that in explored in Özel et al. (2016)).

Other works which suggest stronger constraints always employ assumptions which we have relaxed. Hebeler et al. (2013) and Steiner et al. (2016) have shown that neutron star radii are between 11 and 13 km, assuming that chiral effective theory approaches to neutron matter can be employed above the nuclear saturation density. However, it is difficult to fully quantify the uncertainties in chiral effective theory above the saturation density. The same difficulty is found in the quantum Monte Carlo model we have used above, but we do not employ it at high densities. While Lattimer & Steiner (2014b) ruled out radii larger than 13 km from a similar set of neutron stars, our updated data set and more complete consideration of distance uncertainties weakens the case for smaller radii, as can be seen by comparing our Fig. 3 with Figure 5 in Lattimer & Steiner (2014b) (particularly the change in the constraints for the neutron star in NGC 6304). Guillot et al. (2013) and Özel et al. (2016) have found smaller radii by assuming that (all or most) quiescent LMXB atmospheres must be composed of H. Özel et al. (2016) also use data from photospheric expansion X-ray bursts, but their small radii were strongly driven by the quiescent LMXB data.

Further progress in our understanding of neutron star structure will come from more data which constrains neutron star masses and radii, including additional quiescent LMXB spectra, constraints from NICER and LIGO, and from future missions such as ATHENA, Lynx, and/or STROBE-X. This work has shown that one can approximately quantify the effect that hotspots, atmosphere composition, assumptions about the mass distribution and

<table>
<thead>
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<th>Model</th>
<th>Lower limits 95%</th>
<th>Most prob. 68%</th>
<th>Upper limits 95%</th>
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<td></td>
<td></td>
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<td>11.79</td>
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Constraining the Mass and Radius of Neutron Stars

Table 7. The constraints on the radius, central energy density, and central baryon density of the maximum mass star for the nine scenarios explored in this work.

<table>
<thead>
<tr>
<th>Model</th>
<th>Lower limits 95%</th>
<th>68%</th>
<th>Most prob. 68%</th>
<th>Upper limits 95%</th>
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<td>95%</td>
<td>68%</td>
<td></td>
<td>68%</td>
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<tr>
<td></td>
<td>R_{\text{max}} (km)</td>
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<td>9.891</td>
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<td>11.23</td>
<td>12.31</td>
</tr>
<tr>
<td>90% H</td>
<td>9.720</td>
<td>9.956</td>
<td>10.38</td>
<td>11.21</td>
</tr>
</tbody>
</table>

| ε_{\text{max}} (MeV fm^{3}) |
| baseline | 858.9 | 1153 | 1453 | 1565 | 1631 |
| with X5  | 933.2 | 1210 | 1504 | 1574 | 1642 |
| Model C  | 977.8 | 984.9 | 1000 | 1554 | 1704 |
| H atm.   | 1110  | 1305 | 1508 | 1586 | 1670 |
| M_{\text{max}} > 2.3 | 743.4 | 873.7 | 935.2 | 1129 | 1232 |
| 1.3 < M < 1.5 | 751.3 | 1141 | 1439 | 1586 | 1638 |
| 1.3 < M < 1.7 | 752.9 | 1135 | 1509 | 1622 | 1641 |
| hotspot  | 779.1 | 959.1 | 1214 | 1405 | 1562 |
| 90% H    | 949.0 | 1212 | 1500 | 1574 | 1644 |

| n_{B,max} (fm^{-3}) |
| baseline | 0.7079 | 0.9224 | 1.075 | 1.198 | 1.246 |
| with X5  | 0.7956 | 0.9795 | 1.169 | 1.207 | 1.253 |
| Model C  | 0.8341 | 0.8453 | 0.8687 | 1.263 | 1.300 |
| H atm.   | 0.9120 | 1.044 | 1.165 | 1.221 | 1.268 |
| M_{\text{max}} > 2.3 | 0.6140 | 0.6983 | 0.8581 | 0.8896 | 0.9560 |
| 1.3 < M < 1.5 | 0.6296 | 0.9243 | 1.137 | 1.230 | 1.254 |
| 1.3 < M < 1.7 | 0.6168 | 0.9252 | 1.162 | 1.239 | 1.258 |
| hotspot  | 0.6509 | 0.7834 | 0.9074 | 1.086 | 1.198 |
| 90% H    | 0.7977 | 0.9841 | 1.162 | 1.212 | 1.259 |

Table 7. The constraints on the radius, central energy density, and central baryon density of the maximum mass star for the nine scenarios explored in this work.

the maximum mass, and assumptions about the presence of strong phase transitions have on neutron star radii and the equation of state.

Acknowledgements

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