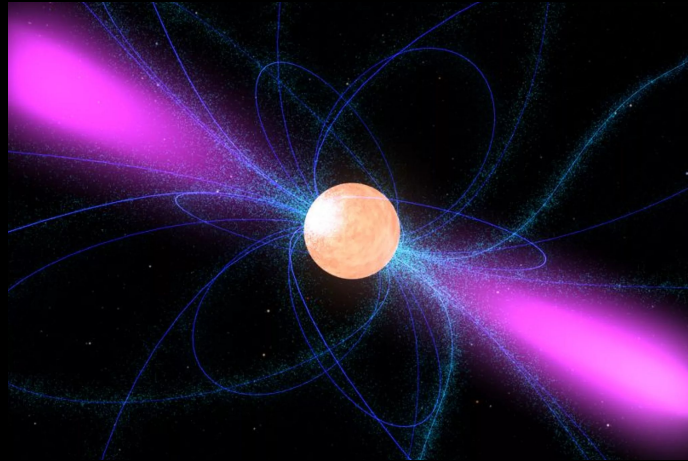


Simulation-based inference for pulsar-population synthesis

Celsa Pardo Araujo

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Credits: NASA

in collaboration with
Michele Ronchi

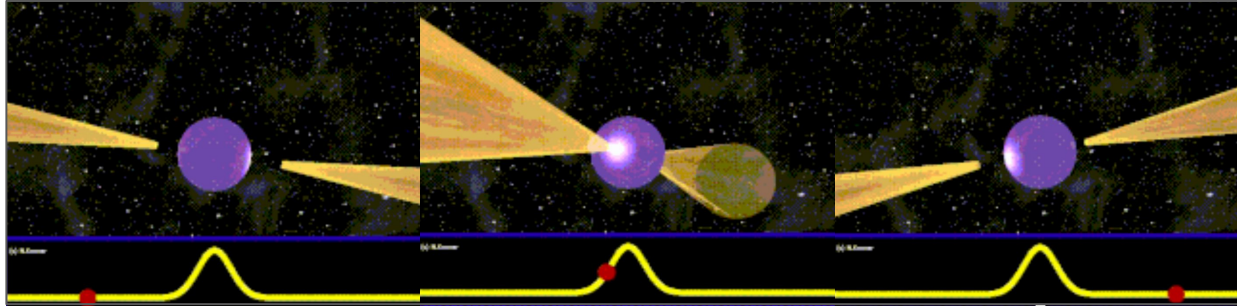


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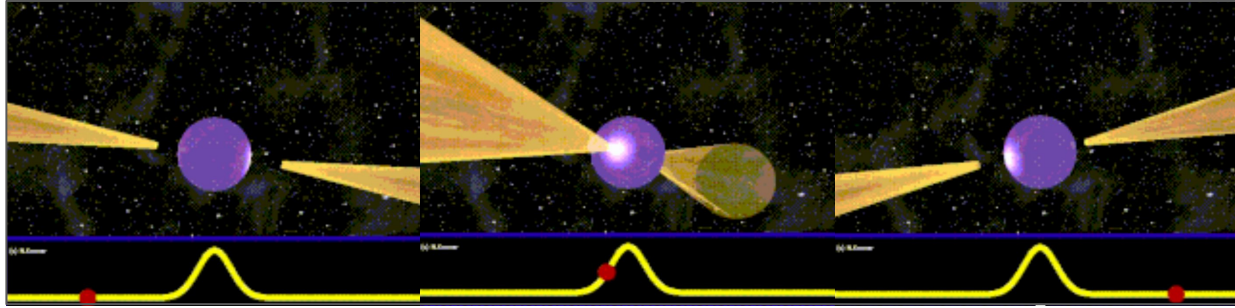


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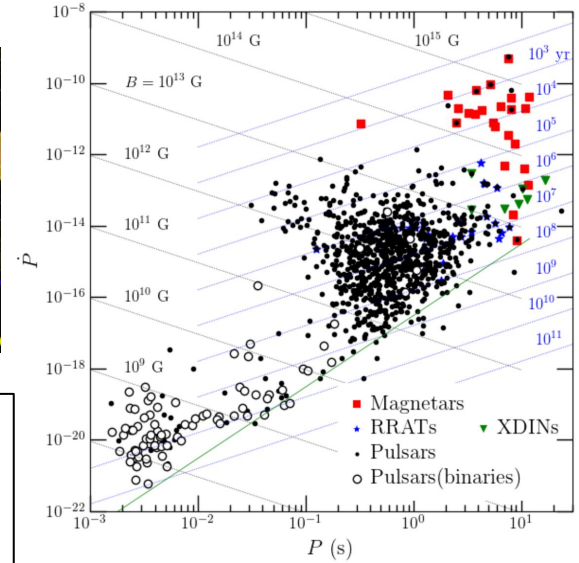
Neutron star population



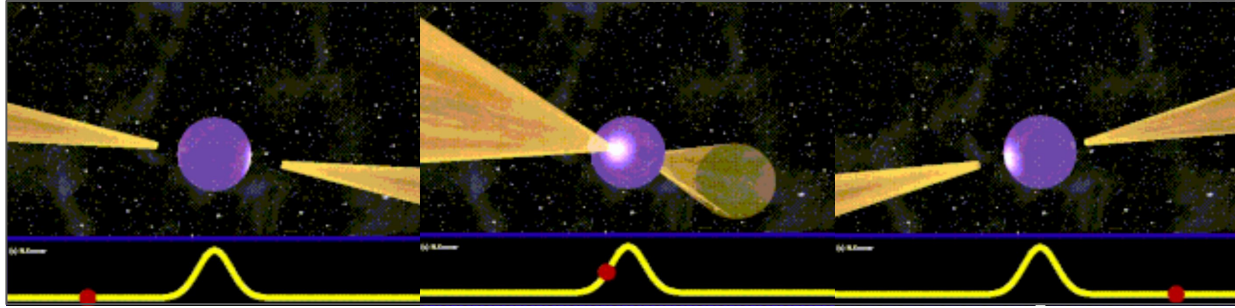
Neutron star population



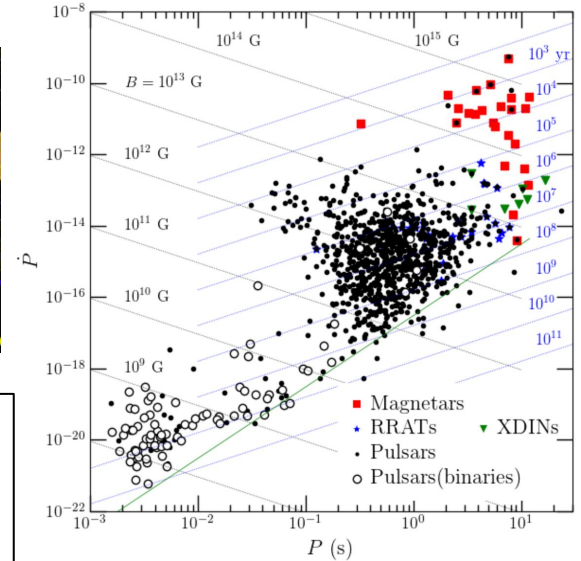
- Different classes of neutron stars occupy different locations in the $P\dot{P}$ -plane probing different evolution paths and/or different origins.



Neutron star population



- Different classes of neutron stars occupy different locations in the $P\dot{P}$ -plane probing different evolution paths and/or different origins.



CC supernova rate:
~ 2 per century

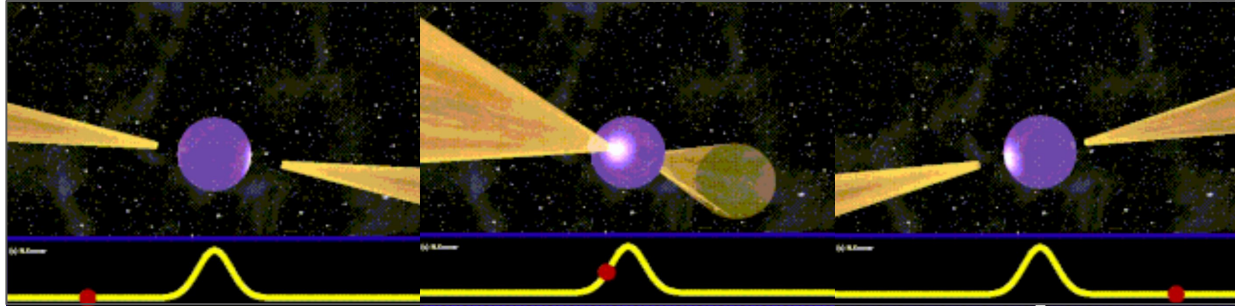
×

Galaxy age:
~ 13.6 billion years

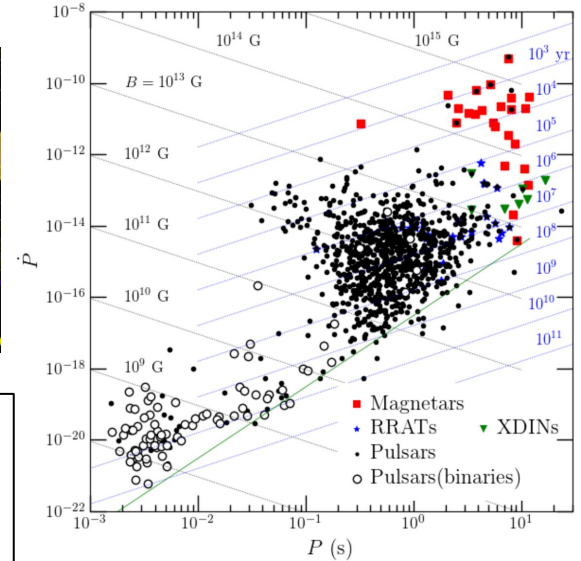
=

NS number:
~ 2.8×10^8

Neutron star population



- Different classes of neutron stars occupy different locations in the P - P -plane probing different evolution paths and/or different origins.



CC supernova rate:
~ 2 per century

×

Galaxy age:
~ 13.6 billion years

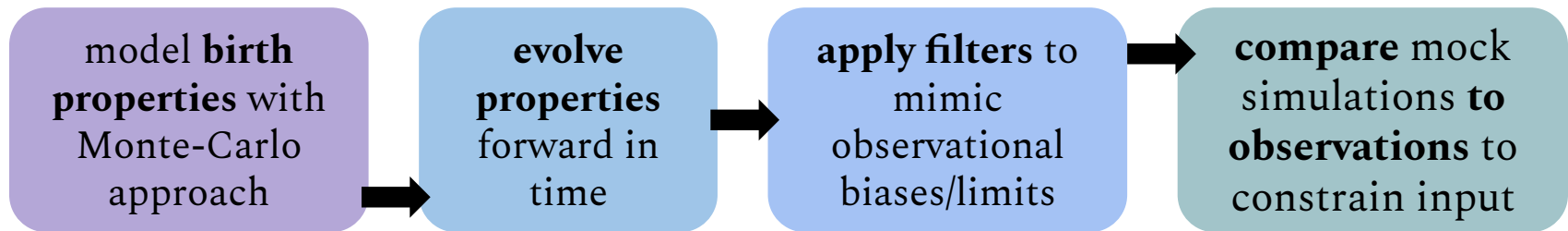
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**~ 3,000
detected
neutron stars!**

- What are the **natal properties** and the **birth rate** of the neutron-star population?
- Are there any **evolutionary links between different neutron-star classes**?

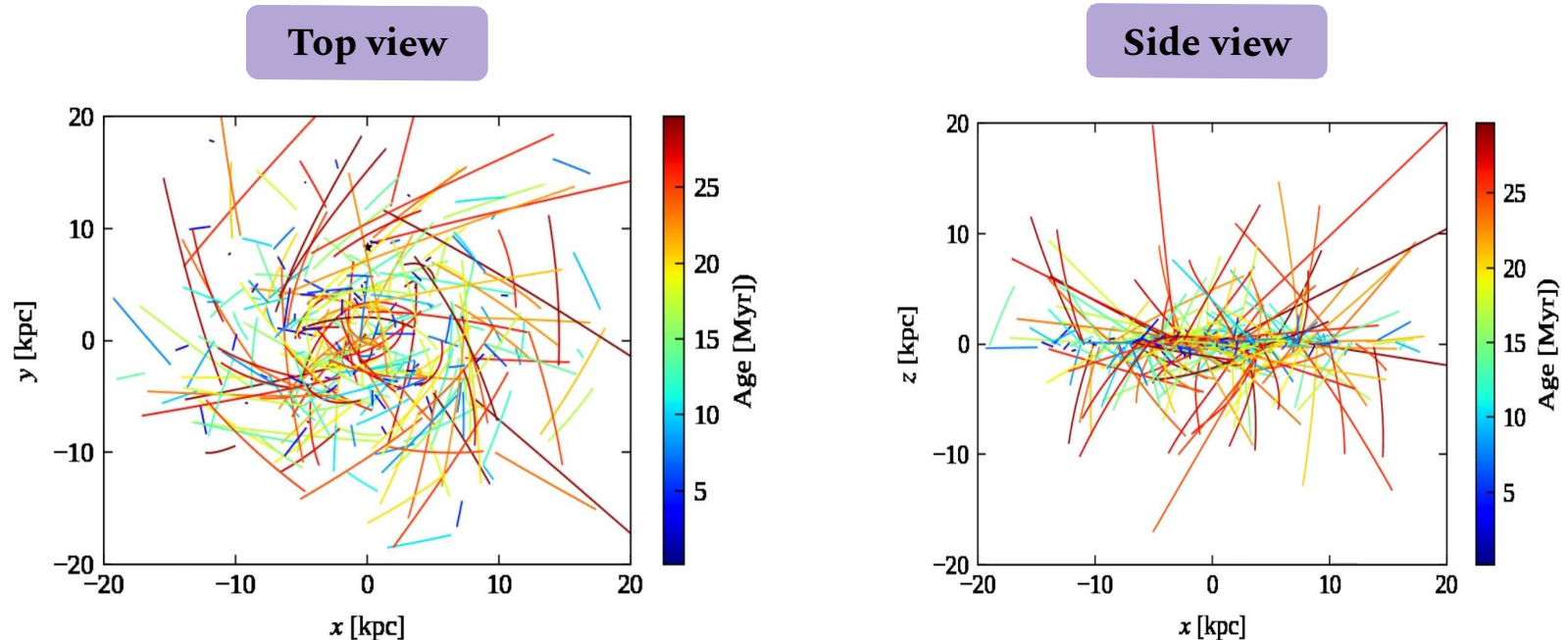
Population synthesis

- What are the **natal properties** and the **birth rate** of the neutron-star population?
 - Are there any **evolutionary links between different neutron-star classes**?
-
- Population synthesis focuses on the **entire** population of neutron stars (e.g. [Faucher-Giguère & Kaspi 2006](#), [Lorimer et al. 2006](#), [Gullón et al. 2014](#), [Ciešlar et al. 2020](#)):



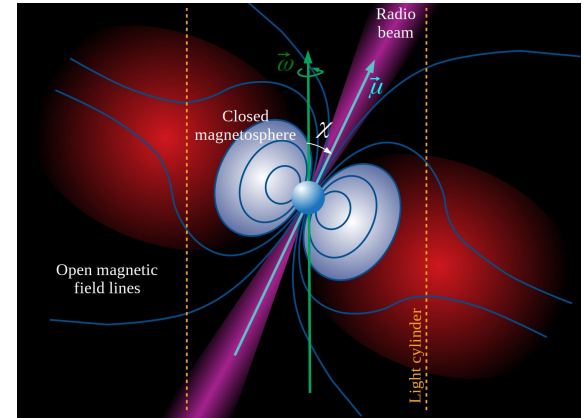
Dynamical evolution

- We evolve the stars' position & velocity by **solving Newtonian equations of motion** in cylindrical galactocentric coordinates: $\ddot{\vec{r}} = -\nabla\Phi_{\text{MW}}$



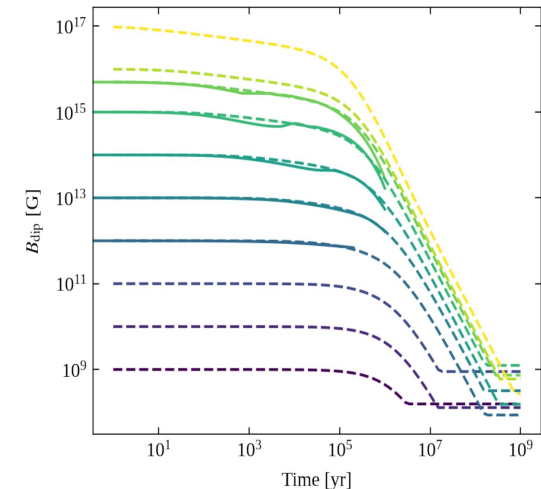
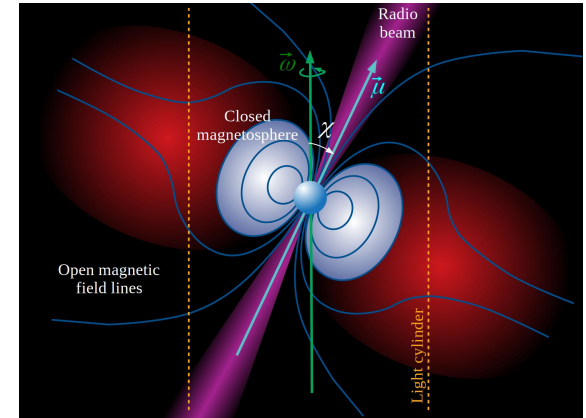
Magneto-rotational evolution

- The neutron-star magnetosphere exerts a **torque onto the star**. This causes **spin-down** and **alignment of the magnetic and rotation axes** (Spitkovsky 2006, Philippov et al. 2014).



Magneto-rotational evolution

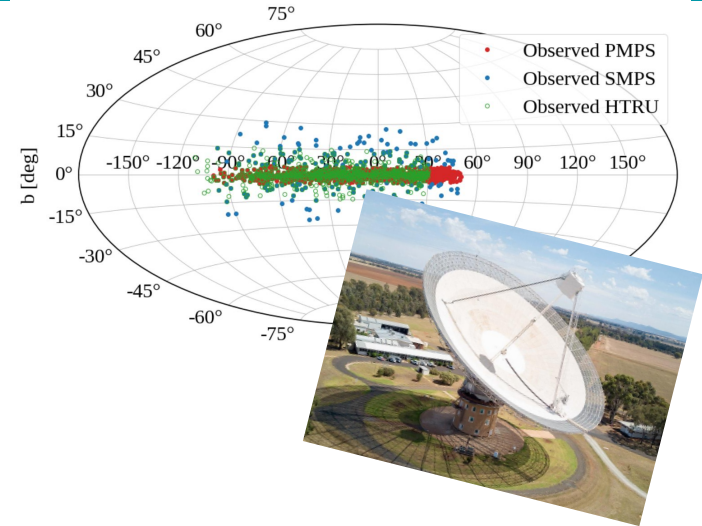
- The neutron-star magnetosphere exerts a **torque onto the star**. This causes **spin-down** and **alignment of the magnetic and rotation axes** (Spitkovsky 2006, Philippov et al. 2014).
- Neutron star **magnetic fields decay** due to the Hall effect and Ohmic dissipation in the outer stellar layer (crust) (e.g., Viganó et al. 2013 & 2021, Gourgouliatos et al. 2014, De Grandis et al. 2020).



Radio surveys & Goals

Radio surveys:

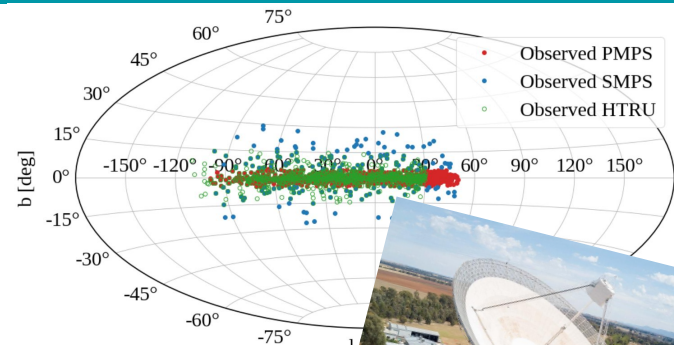
- **Parkes Multibeam Pulsar Survey (PMPS)**
- **Swinburne Parkes Multibeam Pulsar Survey (SMPS)**
- **High Time Resolution Universe Survey (HTRU)**



Radio surveys & Goals

Radio surveys:

- **Parkes Multibeam Pulsar Survey (PMPS)**
- **Swinburne Parkes Multibeam Pulsar Survey (SMPS)**
- **High Time Resolution Universe Survey (HTRU)**



We want to infer:

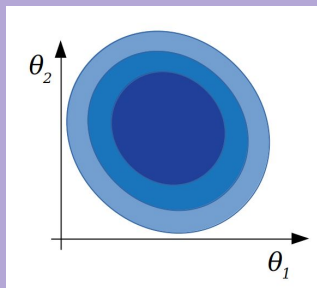
- Period distribution at birth.
- Magnetic field distribution at birth.

Assumptions

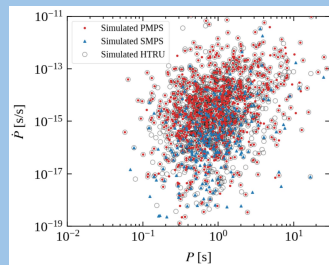
- **Initial periods** follow a normal in log with $\mu_{\log P}$ and $\sigma_{\log P}$ (Igoshev et al. 2022)
- **Initial fields** follow a normal in log with $\mu_{\log B}$ and $\sigma_{\log B}$ (Gullón et al. 2014)

Simulation based inference

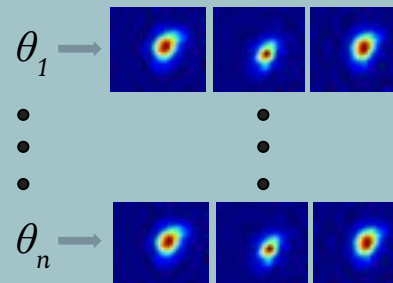
1. **sample** θ_i from the prior $\pi(\theta)$, for $i = 1, \dots, N$



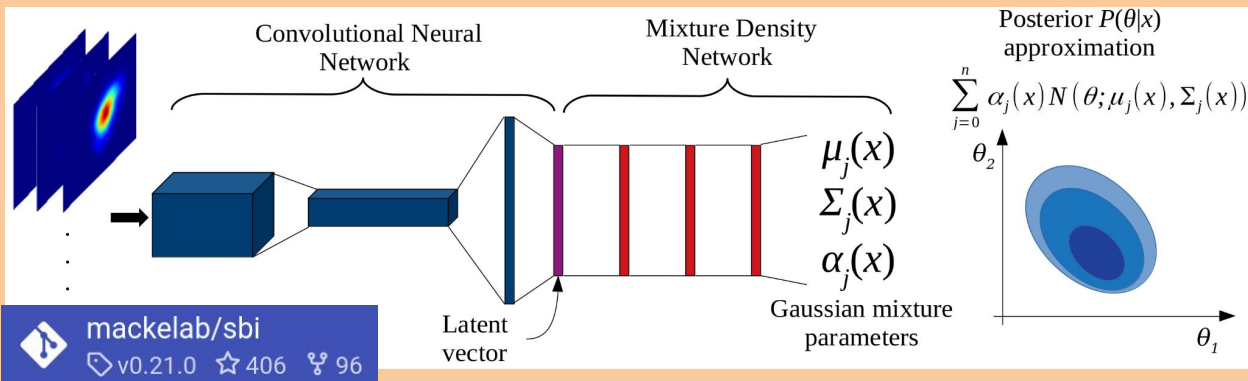
2. **run simulator** for θ_i to produce mock observations: $x_i \sim P(x|\theta_i)$



3. Generate the **training dataset**:

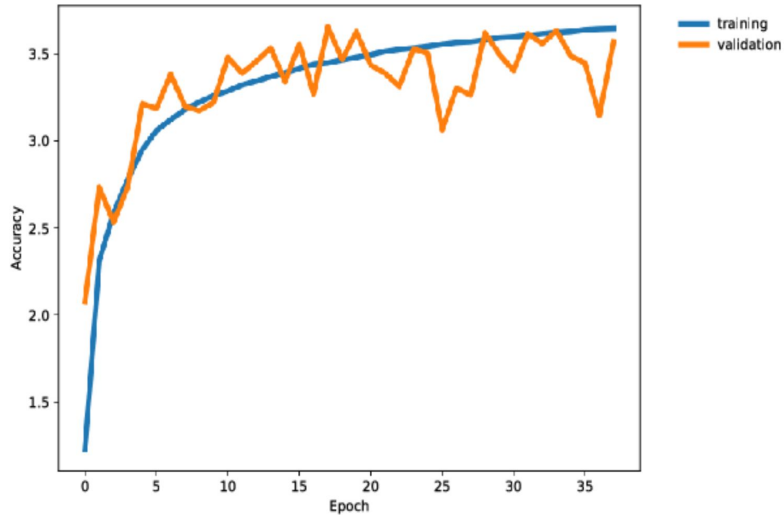


neural network (conditional density estimator) on simulated data to approximate the posterior



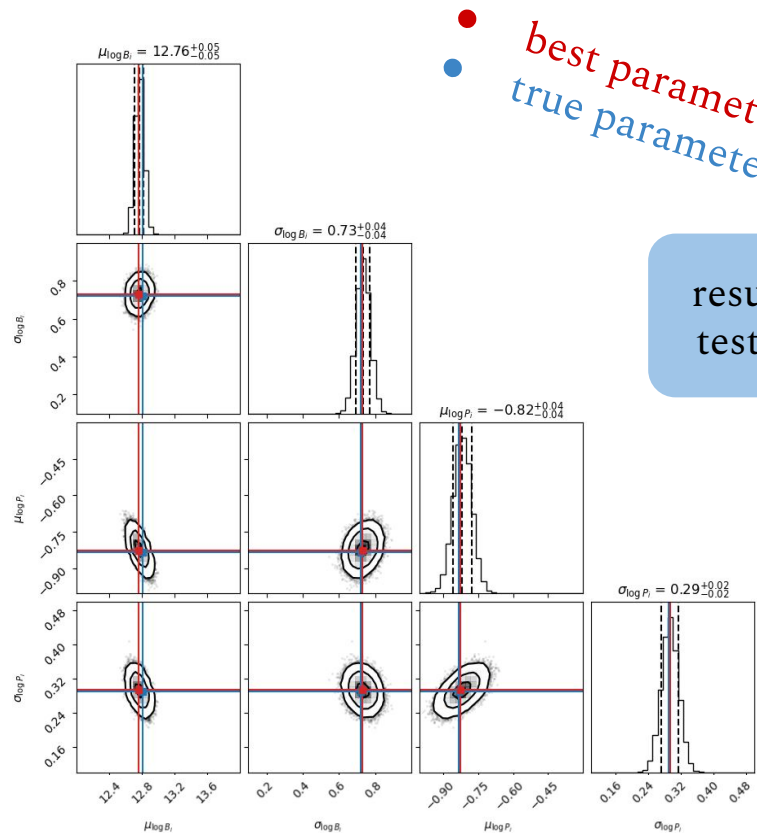
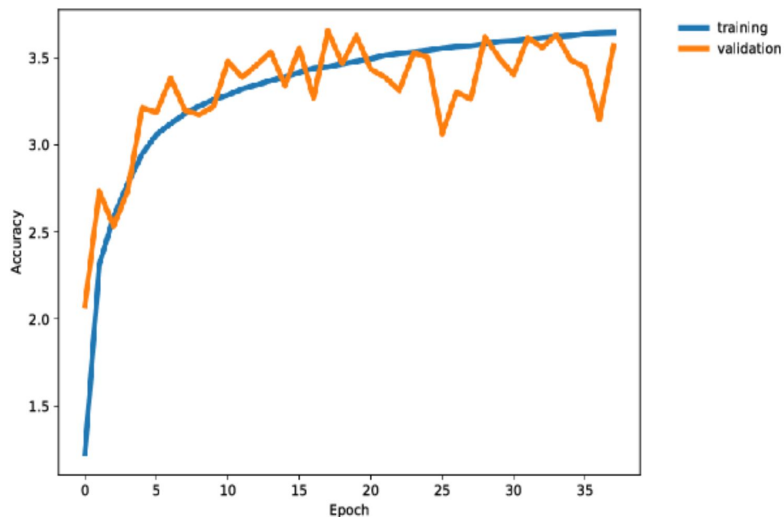
Results

- 360,000 simulations in total, we use 80 % for training, 10 % for validation, 10 % for testing.



Results

- 360,000 simulations in total, we use 80 % for training, 10 % for validation, 10 % for testing.



● best parameters
● true parameters

results for a test sample

Results for the observed population

Preliminary!

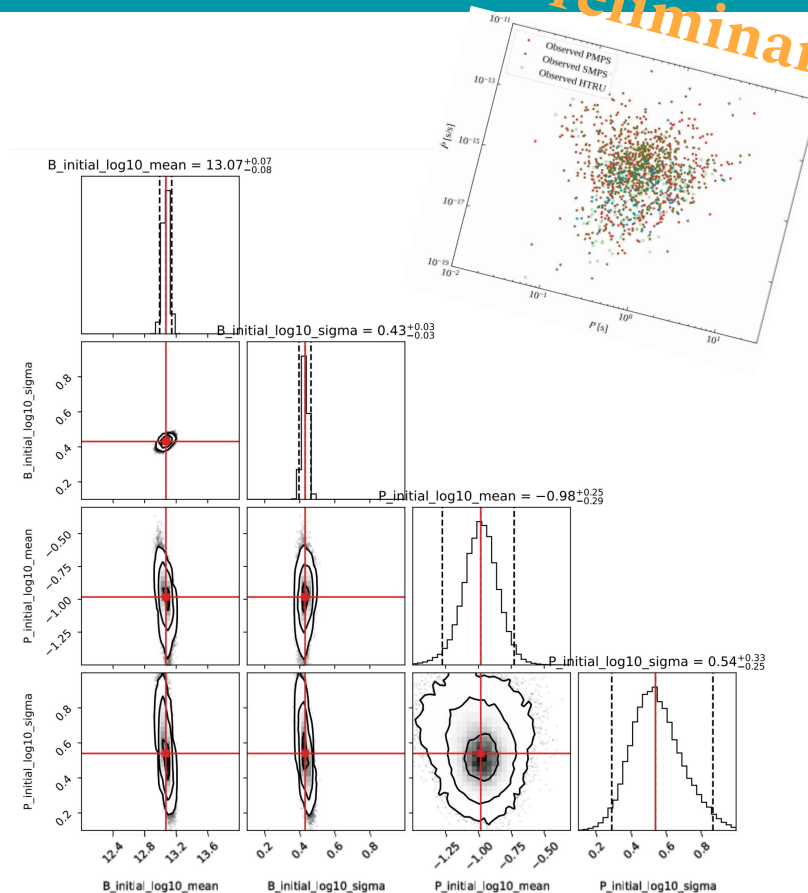
With our optimised neural network, we can also **infer the posteriors** for the **pulsar population detected in our three surveys** and recover the following constraints:

$$\mu_{\log B} = 13.07^{+0.07}_{-0.08}$$

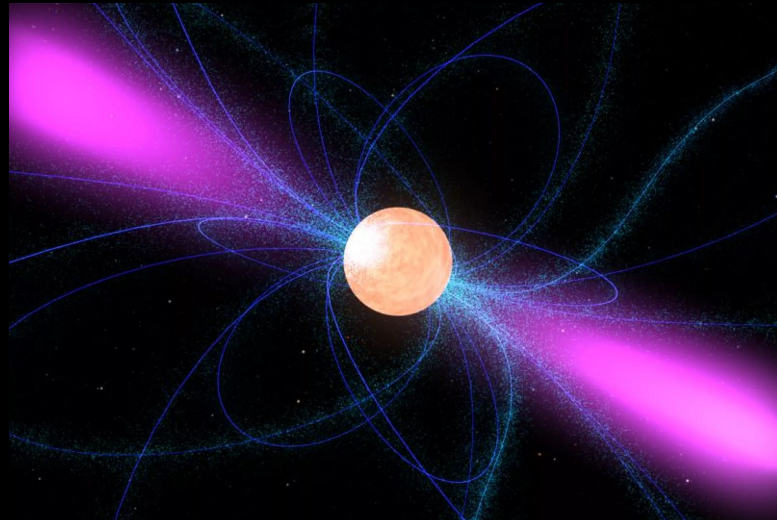
$$\sigma_{\log B} = 0.43^{+0.03}_{-0.03}$$

$$\mu_{\log P} = -0.98^{+0.25}_{-0.29}$$

$$\sigma_{\log P} = 0.54^{+0.33}_{-0.25}$$



Thanks for listening!



Credits: NASA

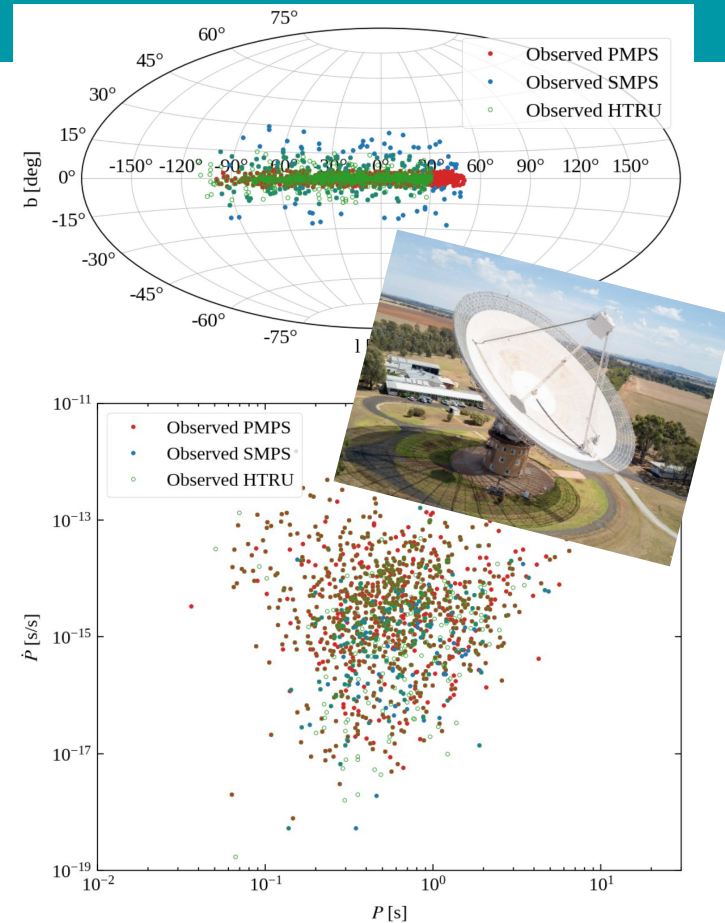


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Radio surveys

- We compare our simulated populations with three surveys from Murriyang (the Parkes Radio Telescope):
 - **Parkes Multibeam Pulsar Survey (PMPS)** (Manchester et al. 2001, Lorimer et al. 2006):
1,009 isolated pulsars
 - **Swinburne Parkes Multibeam Pulsar Survey (SMPS)** (Edwards et al. 2001, Jacoby et al. 2009):
218 isolated pulsars
 - **High Time Resolution Universe Survey (HTRU)** (Keith et al. 2018):
1,023 isolated pulsars



Radio emission and detection

- The stars' **rotational energy** E_{rot} is converted into coherent radio emission (Faucher-Giguère & Kaspi 2006; Gullón et al. 2014).

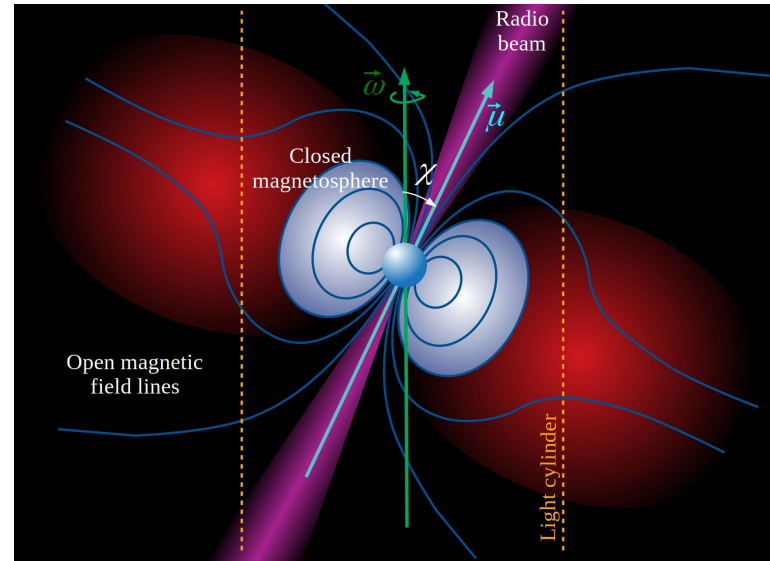
$$L_{\text{radio}} = L_0 \left(\frac{\dot{P}}{P^3} \right)^{1/2} \propto \dot{E}_{\text{rot}}^{1/2}$$

- As **emission is beamed**, $\sim 90\%$ of pulsars do not point towards us. For those intercepting our line of sight, compute **radio flux** S_{radio} & **pulse width** W .

$$S_{\text{radio}} = \frac{L_{\text{radio}}}{\Omega_{\text{beam}} d^2}$$

- A **signal-to-noise ratio** can be estimated through the **radiometer equation**.

$$S/N = \frac{S_{\text{mean}} G \sqrt{N_{\text{pol}} \Delta\nu \Delta t_{\text{obs}}}}{\beta (T_{\text{rec}} + T_{\text{sky}}(l, b))} \sqrt{\frac{P - w_{\text{eff}}}{w_{\text{eff}}}}$$



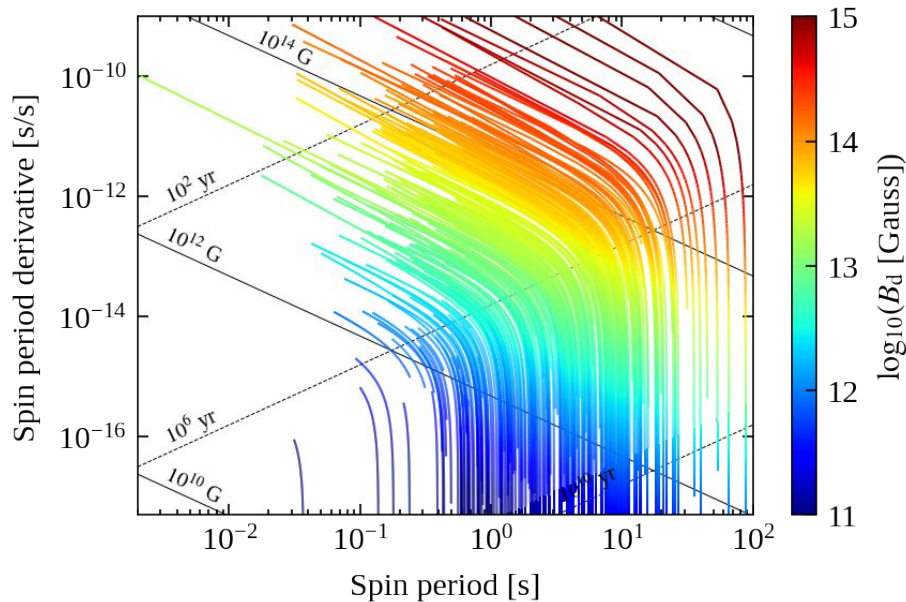
A pulsar counts as detected, if it **exceeds the sensitivity threshold** for a survey recorded with a specific radio telescope.

Magneto-rotational evolution

$$\dot{P} = \frac{\pi^2 B^2 R^6}{c^3 IP} (\kappa_0 + \kappa_1 \sin^2 \chi)$$

$$\dot{\chi} = -\frac{\pi^2 B^2 R^6}{c^3 IP^2} (\kappa_2 \sin \chi \cos \chi)$$

(Spitkovsky 2006,
Philippov et al. 2014)



\dot{P} evolution tracks for

$$\mu_{\log P} = -0.6, \sigma_{\log P} = 0.3, \mu_{\log B} = 13.25, \sigma_{\log B} = 0.75 \text{ and } \alpha = -2.0$$