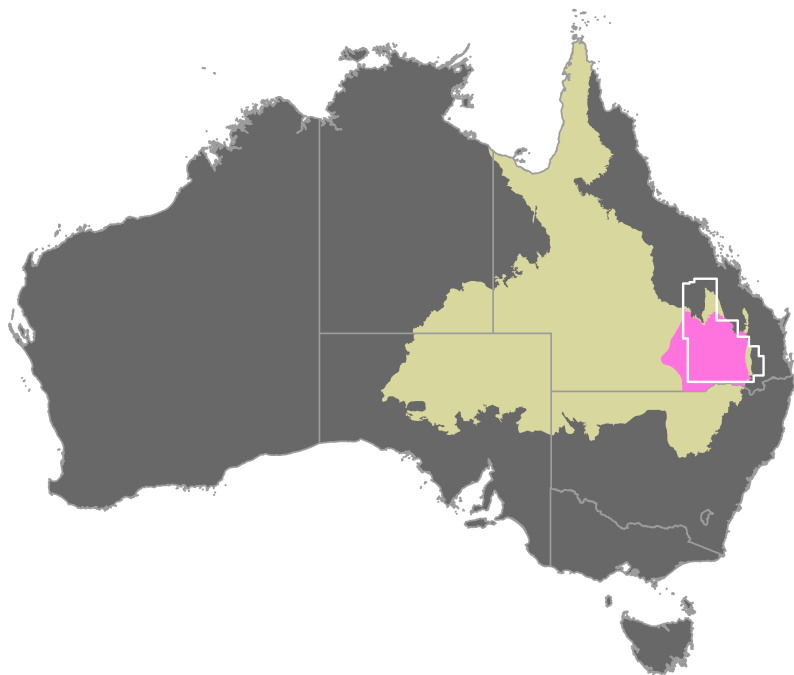




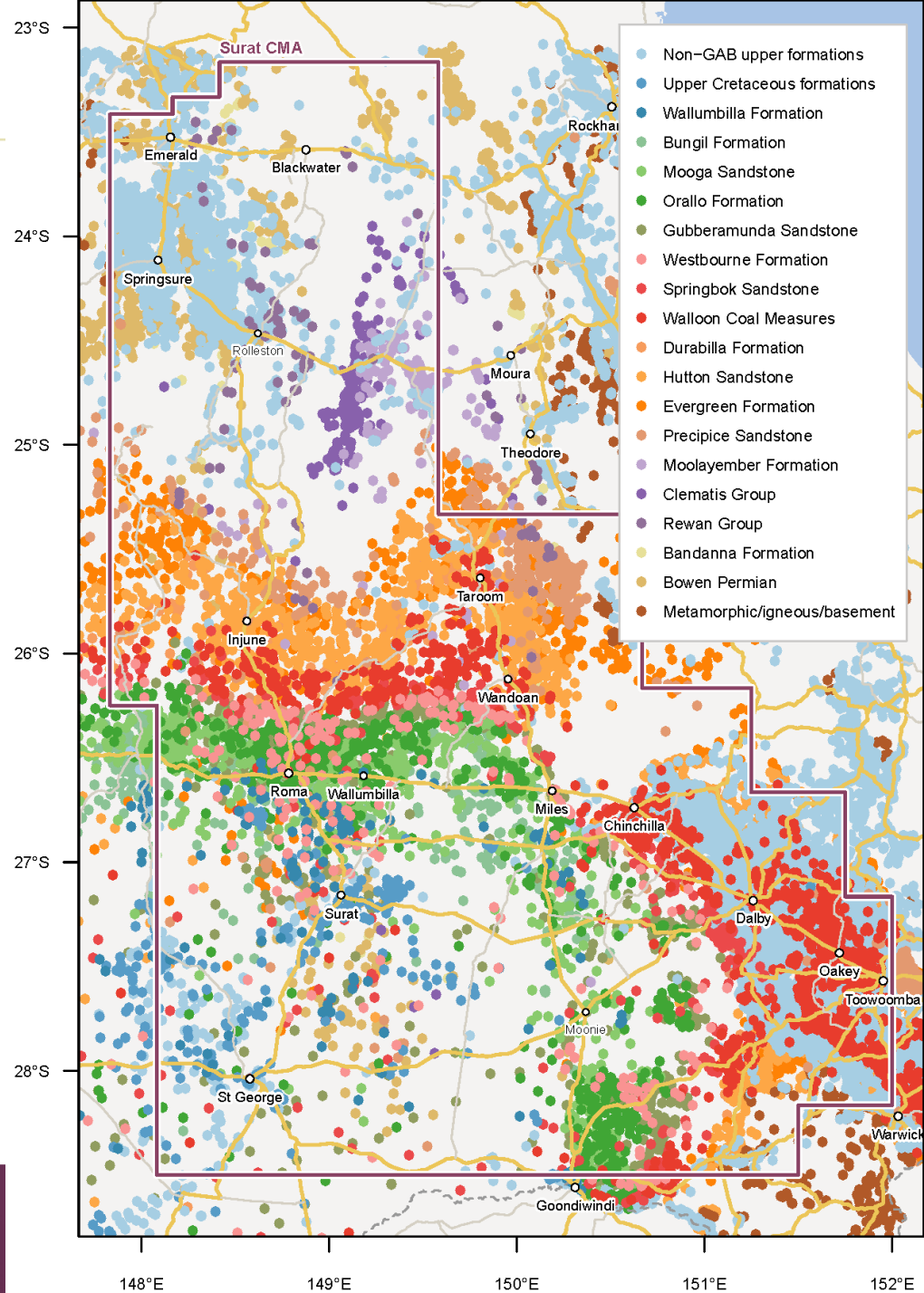
# Monitoring and modelling groundwater extractions over a data-sparse region of Australia

Greg Keir, Nevenka Bulovic, Neil McIntyre

# The Surat Basin



Area – 160,000 km<sup>2</sup>  
 Average rainfall –  
 400 to 800 mm/year

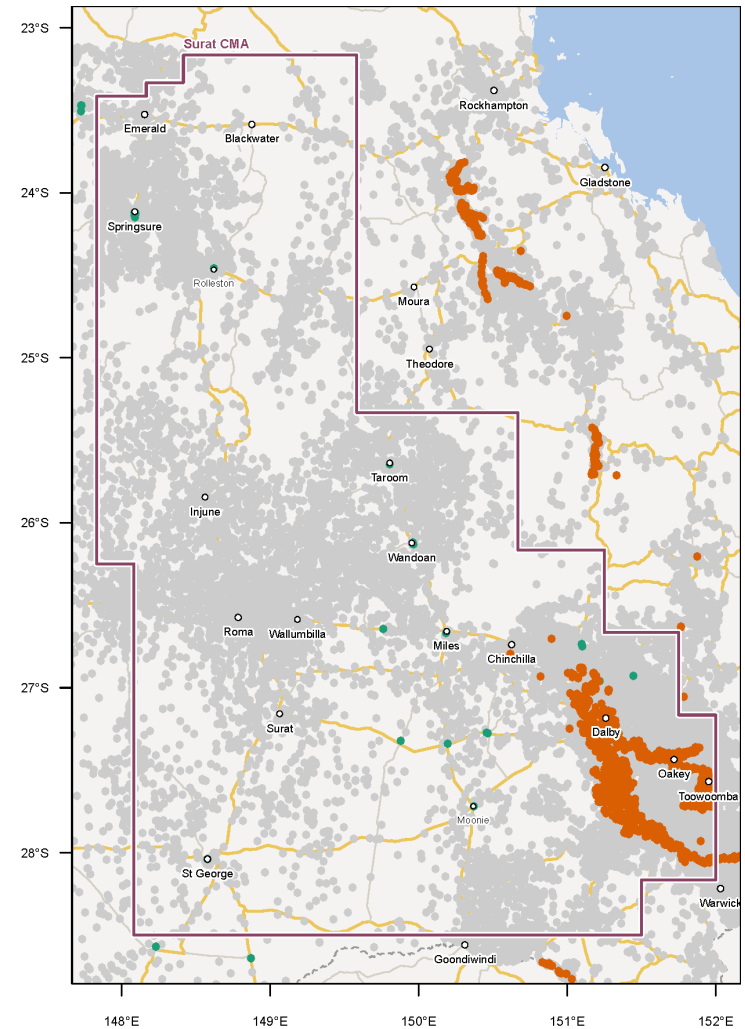






# Data availability – Groundwater extraction metering

- Region has 30,000+ bores
- **< 5000 bores** metered, and metering limited to:
  - Irrigation bores
  - Municipal bores
  - Other larger users
- Most bores are used for stock watering and domestic purposes (S & D)





# Data availability – Stock and domestic bores

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- ‘Direct’ estimates of extraction with different levels of uncertainty
  - Small-scale metering program across 43 bores
  - Anecdotal baseline estimates
  - Long-term analytical and empirical estimates (e.g. OGIA)
- Other supporting (explanatory) datasets that can influence extraction
  - Bore characteristics
  - Hydrogeology and hydrology
  - Climate
  - Rural property characteristics



# Modelling overview

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- A three-step modelling approach used to get bore scale estimates
  1. Magnitude model at property scale
  2. Occurrence model at bore scale
  3. Proportional distribution model within properties

$$g(\mu) = \eta = \sum d\beta$$

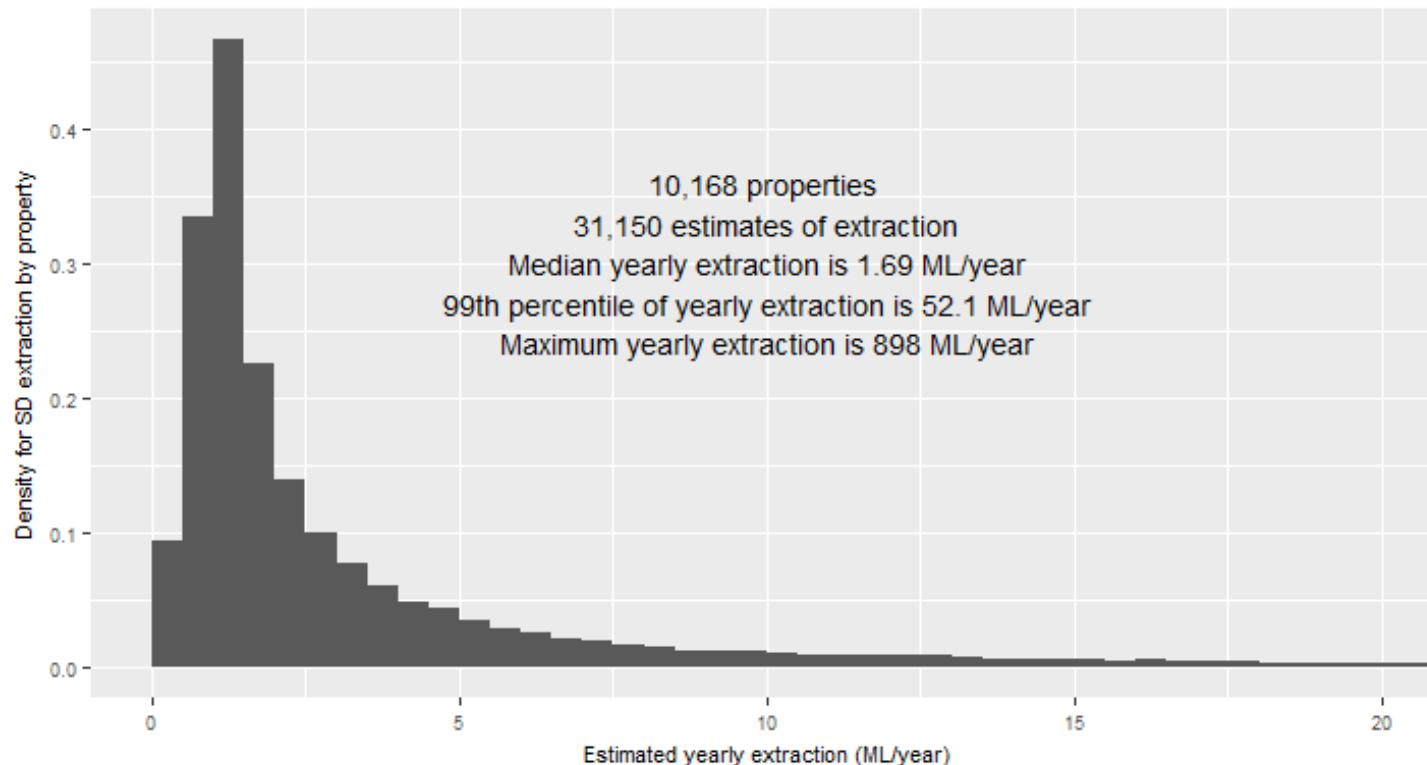




# Property magnitude model

- Response variable: Gamma distribution
- Inverse link function:  $g(\mu_i) = \frac{1}{\mu_i} = \sum_{j=1}^p d_{i,j} \beta_j$

extraction  $\sim$  rain + temp\_max + temp\_min + solar + ndvi +  
surf\_water + Persons + Area + Pasture + GIS\_LNG +  
GIS\_LAT + nBores



# Occurrence model

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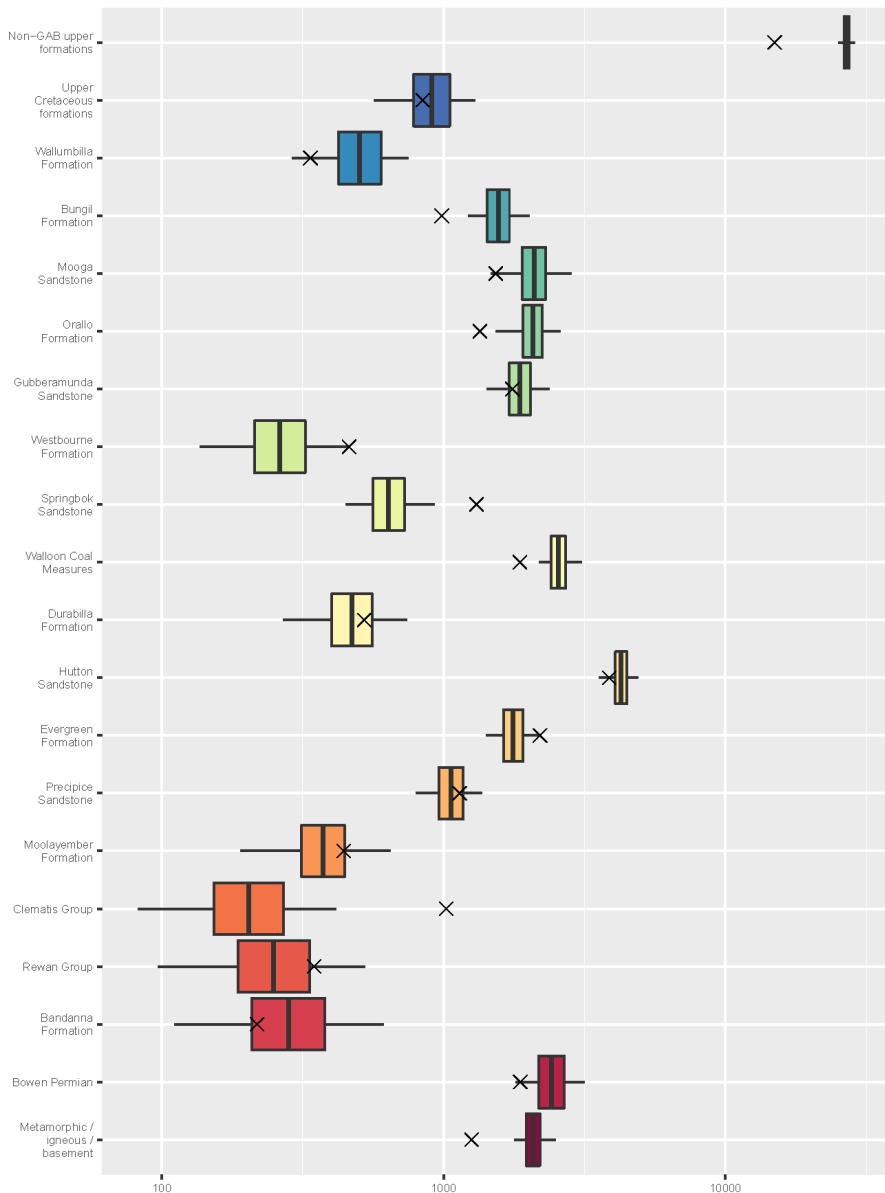
- Response variable: Bernoulli distribution
- Logit link function:  $g(\pi_i) = \log\left(\frac{\pi_i}{1-\pi_i}\right) = \sum_{j=1}^p d_{i,j}\beta_j$

Used ~ FACILITY\_TYPE + DN + *well* + Age + GIS\_LNG + GIS\_LAT +  
EC + Deep\_Aqui + Thickness + *t* + *tb* +  
*rain* + *temp\_max* + *temp\_min* + *solar* + *ndvi* +  
*surf\_water* + Persons + Area + Pasture

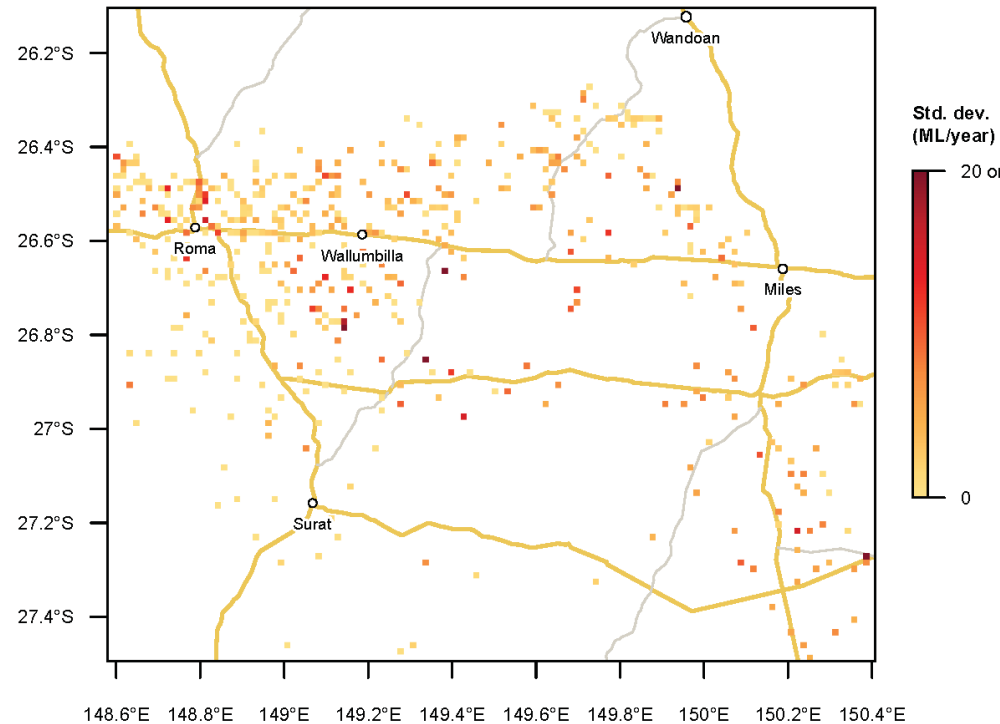




# Example outputs



Total extraction by aquifer



Spatial distribution of extraction

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