### Approximate Bayesian Computation for Planet Occurrence Rate

Hsu, D. C. et al., 2018, The Astronomical Journal, 155:205.

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## Introduction

Kepler transit mission



x 2700 exoplanets

Does *Kepler* data directly indicate the exoplanet population?

- Orbital period
- Eccentricity
- Host-planet size ratio
- Signal to noise
- Target selection

Biases on the transit probability

Discovered exoplanets via transit by *Kepler* 

Approximate Bayesian Computation

True exoplanet population

NO!

### ABC method

First idea from Rubin, D. B. (1984) Named and established by Beaumont, M. A. et al. (2002)

Standard Bayesian :

$$P(A|B) = \frac{P(B|A)P(A)}{P(B)}$$

Posterior <</p> **Likelihood** × Prior

How to approximate likelihood function?

1) Simulate an artificial data with the arbitrary model parameters.

- 2) If 1) is close to the observed data, it is an acceptable model.
- 3) Accumulate acceptable models by repeating 1) 2).
- 4) The final distribution of 3) indicates the model likelihoods.

Start ABC method gif (Leyshon, 2021)

### ABC method

Key parameters for the ABC method:

- Summary statistics of the observed data  $(S(Y_{obs}))$
- Summary statistics of the artificial data (S(Y))
- Array of model parameters  $(\phi)$
- Distance function ( $\rho$ )
- Distance threshold ( $\epsilon$ )

$$\pi_{ABC}(\phi|S) = \pi(\phi|\rho(S(Y), S(Y_{obs})) < \epsilon)$$

Application of Monte-Carlo simulation and Importance sampling:

- Sampling model parameter sets with a Monte-Carlo simulation provides S(Y).
- *ϵ* can be updated by weighting the distribution of a model parameter from a previous ABC generation (Importance sampling).

### Goal

To find planet occurrence rates over different exoplanet periods and radii.

```
Period = from 0.5 to 320 days
Radius = from 0.5 to 16 R_{Earth}
```

Each bin = (period, radius) is independent.

#### Procedure

- 1) Create a planetary catalogue with a given occurrence rate.
- 2) Simulate transit detection using the planetary catalogue from 1) and create an observed catalogue.
- 3) Derive summary statistics and distance function for each bin.
- 4) Gather a set of acceptable occurrence rates.
- 5) Find ABC posterior and update prior distribution and distance threshold for each bin.
- 6) Repeat 1) 5) until it meets the ending condition.

1) Create a planetary catalogue with a given occurrence rate.

 $p_{r,p} = \text{Prior distribution of occurrence rate per bin.}$   $f_{r,p} = \text{Selected occurrence rate from P}_{r,p} \text{ per bin.}$   $N_{targ} = 150,518 \text{ target host stars from the Kepler data.}$  $n_{s,r,p} = \text{Number of planets per host star } \leftarrow \text{Poison}(k \le 10, \lambda = \sum_{r,p} f_{r,p}).$ 



2) Simulate transit detection using the planetary catalogue from 1) and create an observed catalogue.

	$P = \text{Period}^{(c)}$ $d = \text{Transit depth}^{(c)}$ $D = \text{Transit duration}^{(c)+\{i,e,\omega\}}$			(C (r (t	<ul> <li>(c) value from the catalogue in 1).</li> <li>(p) arbitrary value per planet</li> <li>(t) arbitrary value per transit simulation</li> </ul>		
$t_o = 1$ ime of 0 <sup>err</sup> transit <sup>(i)</sup> i = Inclination <sup>(p)</sup> e = Eccentricity <sup>(p)</sup> $\omega = $ Angular velocity <sup>(p)</sup>							
b = Transit impact parameter <sup>(t)</sup>				er <sup>(t)</sup>	Detectable by Keple		
Relocate planets by $\{\hat{P}, \hat{R}_{p}\}$				, $\hat{R_p}$	YES NC	) ~	
	Planet 2	-	-		Find "measured"		
-		Planet 4	Planet 1 :	: Planet n	$\hat{P}, \hat{d}, \hat{D}, \hat{t}_0$		
	:	Planet 3 :		:	$\hat{R}_{n}$		
Observed catalogue							

3) Derive summary statistics and distance function for each bin.

 $N_{targ} = 150,518$  target host stars from the *Kepler* data.  $N_p = 3380$  planet candidates for  $N_{targ}$  from the *Kepler* data.

 $n_{rn}$  = Number of planets per bin in the observed catalogue.

Planet 2	:	:	:
:	:	:	:
:	Planet 4	Planet 1	:
:	:	:	Planet n
:	Planet 3	:	:

Observed catalogue

 $\mathbf{n}_{r1,p1} \mid \mathbf{n}_{r1,p2} \mid \mathbf{n}_{r1,p3} \mid$ n<sub>r1,p4</sub>  $\begin{array}{|c|c|c|c|c|c|c|c|} n_{r2,p1} & n_{r2,p2} & n_{r2,p3} & n_{r2,p4} \\ \hline n_{r3,p1} & n_{r3,p2} & n_{r3,p3} & n_{r3,p4} \\ \end{array}$ 

Number of observed planets

Summary statistics of a model :

Summary statistics of *Kepler* data :

**Distance function :** 

$$S(Y)_{r,p} = \frac{n_{r,p}}{N_{targ}}$$
$$S(Y_{obs})_{r,p} = \frac{\{N_{(r,p)} \subset N_p\}}{N_{targ}}$$

 $\rho(S(Y_{obs}), S(Y))_{r, \rho} = \{S(Y) - S(Y)\}^2$ 

### 4) Gather a set of acceptable occurrence rates.

 $N_{draw}$ ,  $N_{part}$  = Arbitrary number of trials and survivals.

 $f_i = A \mod occurrence rate (i).$ 

 $\rho_i$  = A distance function of a model occurrence rate (*i*).

 $\epsilon$  = Distance threshold

### At 0<sup>th</sup> generation:



5) Find ABC posterior and update prior distribution and distance threshold for each bin.

 $f_i$ ,  $\rho_i$  = An occurrence rate and its distance function.  $\sigma$  = Standard deviation of acceptable occurrence rates.  $N_{f_{i},\sigma}$  = Probability function of an occurrence rate in a form of normal distribution.  $w_i$  = Normalised sampling weight of an occurrence rate.  $p_{0,r,p}$  = Prior distribution function at 0<sup>th</sup> generation (IDEM).  $p_{r,p}$  = Prior distribution function at current generation.

Sampling weight :

$$w_{i}^{*} = \frac{p_{0,r,p}(f_{i},\sigma)}{p_{r,p}(f_{i},\sigma)} \rightarrow \text{Normalised}: \quad w_{i}^{*} = \frac{W_{i}}{\sum_{i}^{N_{part}} w_{i}^{*}}$$

$$p_{ABC}(f|Y_{obs})_{r,p} = \sum_{i}^{N_{part}} w_{i}N_{f}(f_{i},\sigma)_{r,p}$$

 $\epsilon = max(\rho_i)$ 

ABC posterior :

Prior distribution for next generation :  $p_{(IS)r,p} = \sum_{i}^{p_{ait}} w_i N_f(f_i, \tau \sigma)_{r,p}$ Distance threshold for next generation :

where 
$$\tau \simeq 2$$

### 1

### ABC method for Kepler data

6) Repeat 1) - 5) until it meets the ending condition.

Ending conditions :

(1) Mean distance function is less than the distance threshold.

- (2) The ABC generation reaches to 200.
- (3) Number of same occurrence rate drawing  $> N_{part}$ .
- (4) Median number of  $N_{draw} > 0.2*N_{max}$ .
- (5) No improvement of  $\epsilon$  in consecutive three generations.

The finalised ABC posterior tells the likelihood of the occurrence rate!

Uncertainty of the ABC posterior :

- Reaches to Monte-Carlo noise limit.
- Shorter period bin yields smaller uncertainty.



Figure 5. ABC estimated occurrence rates for the Q1–Q16 planet candidates orbiting FGK stars using the Christiansen et al. (2015) gamma CDF curve for the detection efficiency. The numerical values of the occurrence rates are stated as percentages (i.e.,  $10^{-2}$ ). The color coding of each cell is based on  $(d^2f)/[d(\log_2 R_p) d(\log_2 P)]$ , which provides an occurrence rate normalized to the width of the bin and therefore is not dependent on choice of grid density. Cells colored gray have estimated upper limits for the occurrence rate. Note that the bin sizes are not constant.



Figure 6. Ratios of the planet occurrence rate inferred by the IDEM to the planet occurrence rate inferred by the ABC method based on the Q1–Q16 candidates orbiting FGK stars. The color indicates the relative difference between the occurrence rates scaled by the uncertainty given by the standard deviation from the ABC method. The two methods give similar results for easily detected planets (most bins with high S/N are within  $\sim$ 1 standard deviations), but the IDEM substantially underestimates the frequency of near-Earth-size planets for orbital periods beyond  $\sim$ 80 days.

#### \* $f_{IDEM}$ : Same as the means of the prior distribution at 0<sup>th</sup> generation.

## Summary

**Approximate Bayesian Computation** 

- Monte-Carlo method using a distribution of models.
- The likelihood is approximated by accumulating "good" models.
- A final posterior indicates the likelihood distribution of the models.

Exoplanet occurrence rate from *Kepler* data with ABC application

- The uncertainty of the ABC posterior is as small as the Monte-Carlo noise limit.
- The occurrence rates for Earth-like planets > 80 days period is larger than the former expectation.

### Reference

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