



# Frequentists and Bayesian methods to incorporate recruitment rate stochasticity at the design stage of a clinical trial

Pilar Pastor Martínez

Supervision by Malgorzata Roos

Biostatistics Master Exam



## Content

- Recruitment and Patient Leakage
- Methods for Recruited Counts
- Methods for Waiting Time
- Exact methods vs MC simulations
- Conclusions
- Reproducibility ([GitHub](#))



## Recruitment and Patient Leakage



## Why recruitment rates?

According to [Carter \(2004\)](#)

Timely recruitment vital to the success of a clinical trial

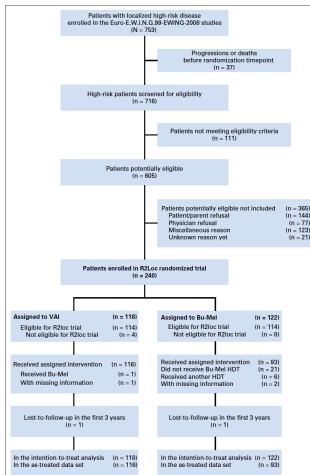
Inadequate number of patients → lack of power

Recruitment period too long → competing treatments

Recruitment of patients varies at each stage

Methods applicable to all the stages

# CONSORT





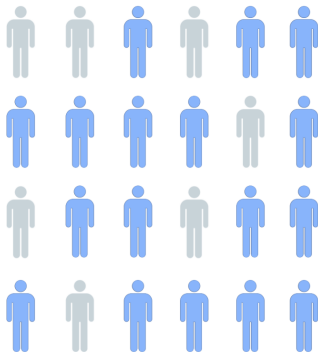
## Target Population



Target Population

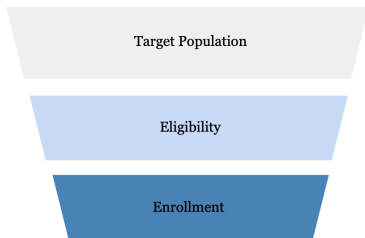
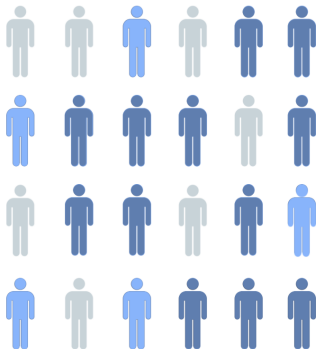


## Eligibility



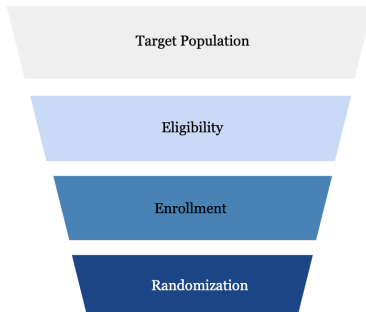
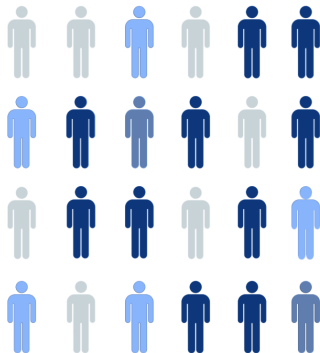


## Enrollment



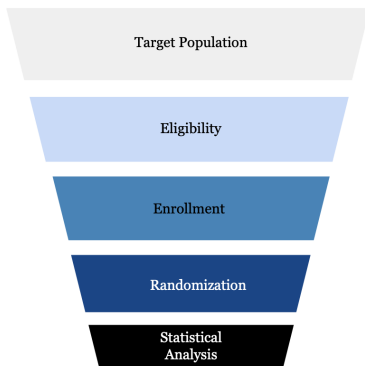
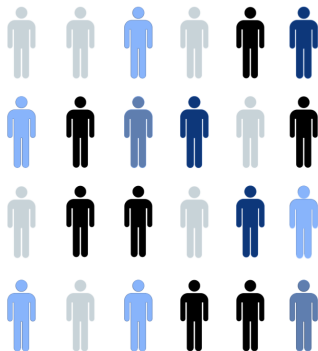


## Randomization

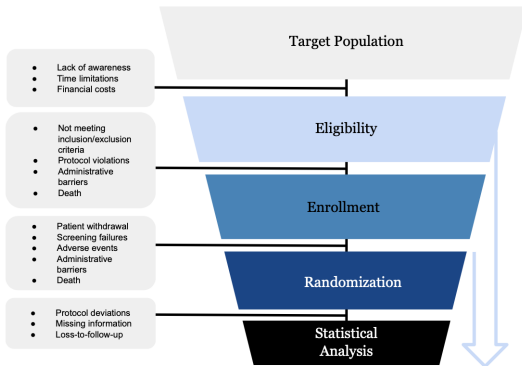




## Statistical Analysis



## Patient Leakage





## Definitions

**Recruitment rate:** Per time-unit ([Piantadosi, 2024](#))

$$\lambda = \frac{\Delta C}{\Delta T} = \frac{C_1 - C_0}{T_1 - T_0} = \frac{C_1}{T_1}$$

**Accrual:** Cumulative Recruitment

**Aleatory uncertainty:** randomness inherent and unpredictable

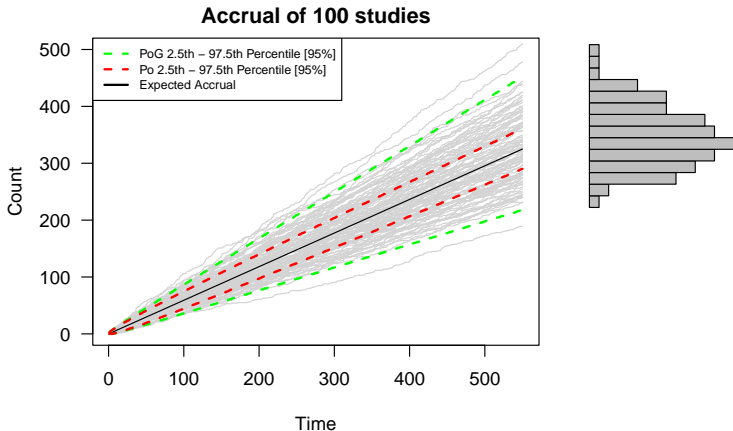
**Epistemic uncertainty:** arises from limited knowledge about recruitment rates



## Methods for Recruited Counts



## Motivation Models for Counts





## Models for Counts

**Recruitment** in unit of time ( $t=1$ ):

Methods	Counts	Expectation	Variance	Aleatory	Epistemic
Expectation	$C = \lambda$	$\lambda$	0	No	No
Poisson	$C \sim \text{Po}(\lambda)$	$\lambda$	$\lambda$	Yes	No
Poisson - Gamma	$C \sim \text{Po}(\Lambda); \Lambda \sim G(\alpha, \beta)$	$\frac{\alpha}{\beta}$	$\frac{\alpha(\beta+1)}{\beta^2}$	Yes	Yes

**Accrual** for time  $t$   $[0, t]$ :

Methods	Counts	Expectation	Variance	Aleatory	Epistemic
Expectation	$C(t) = \lambda t$	$\lambda t$	0	No	No
Poisson	$C(t) \sim \text{Po}(\lambda t)$	$\lambda t$	$\lambda t$	Yes	No
Poisson - Gamma	$C(t) \sim \text{Po}(\Lambda t); \Lambda \sim G(\alpha, \beta)$	$t \frac{\alpha}{\beta}$	$t \frac{\alpha(\beta+t)}{\beta^2}$	Yes	Yes



## Multicenter Trial on Palliation in Terminal Esophageal Cancer

Example from [Carter \(2004\)](#):

Recruitment Rate  $\lambda = 0.591$  per day

Time  $T_{\text{target}} = 550$  days





## Multicenter Trial on Palliation in Terminal Esophageal Cancer

Example from [Carter \(2004\)](#):

Recruitment Rate  $\lambda = 0.591$  per day

Time  $T_{target} = 550$  days

Models for Counts at time point  $t$ :

**Expectation:**  $EC(t) = \lambda t$

**Poisson:**  $C(t) \sim Po(\lambda t)$

**Poisson - Gamma:**  $C(t) \sim Po(\Lambda t); \Lambda \sim G(\alpha, \beta)$

$$\alpha = 32.4 \text{ and } \beta = 54.8$$

$$E\Lambda = \frac{\alpha}{\beta} = 0.591$$



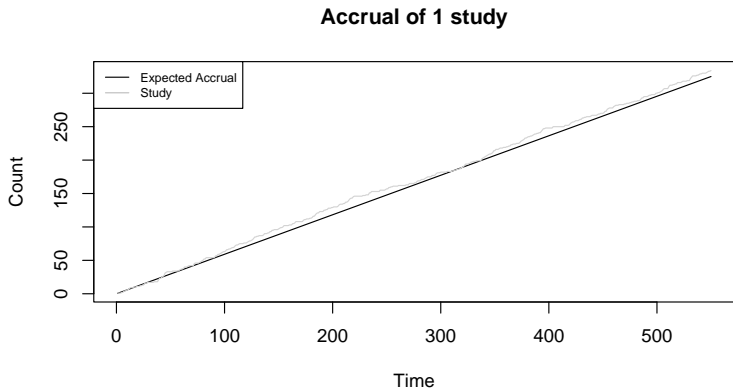
## Accrual at time point $t$

$$\text{Expectation: } EC(t) = E(\underbrace{C + \dots + C}_{t \text{ times}}) = tEC = \lambda t$$

$$\text{Poisson: } \underbrace{Po(\lambda) + \dots + Po(\lambda)}_{t \text{ times}} = Po(\lambda t)$$

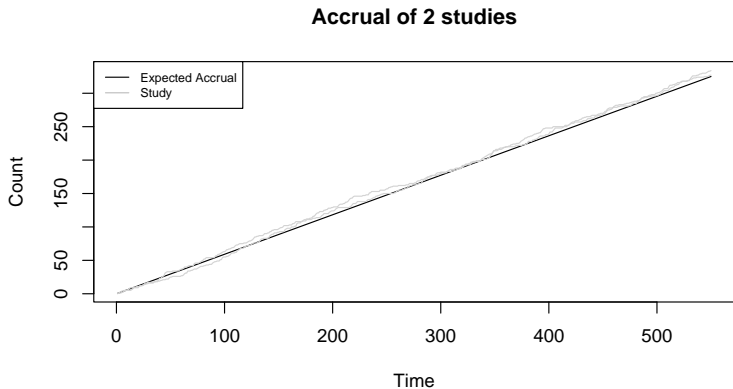


## Accrual of 1 study

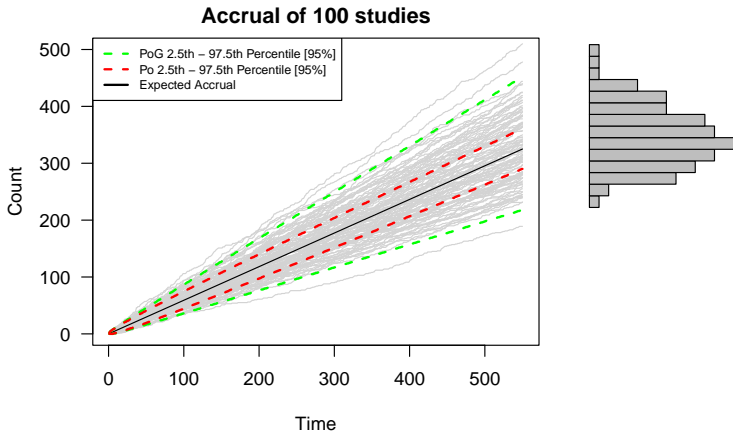




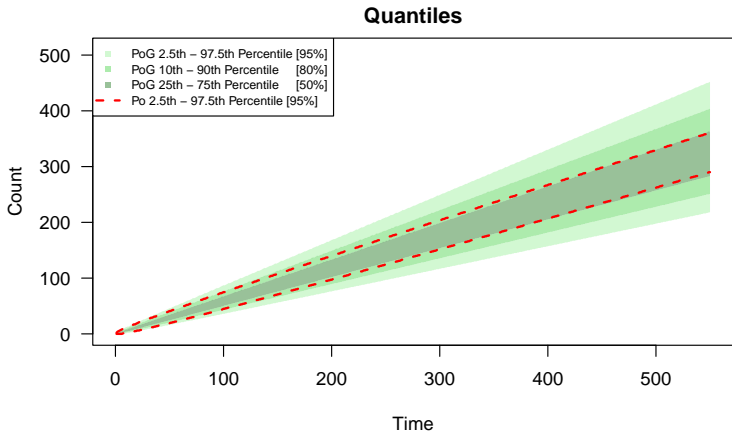
## Accrual of 2 studies



## Accrual of 100 studies



## Exact uncertainty bands





## Two versions of randomness of $\lambda$

**Version 1:** Random recruitment rate realization  $\lambda$  varies across studies and remains **fixed** within study **over time**

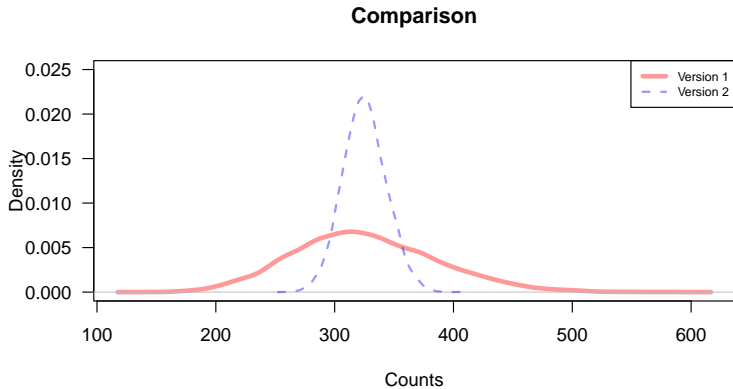
→ PoG distribution

**Version 2:** Random recruitment rate realization  $\lambda$  varies across studies and **varies** within study **over time**

→ Distribution with surprising properties



## Version 1 different from Version 2





## Negative binomial derived from Poisson-Gamma model at time point $t$

Let  $C(t)|\Lambda \sim Po(\Lambda t)$  and  $\Lambda \sim G(\alpha, \beta)$

$$\begin{aligned} p(c) &= \int_0^\infty p(c|\lambda)p(\lambda)d\lambda \\ &= \int_0^\infty \frac{(\lambda t)^c \exp(-\lambda t)}{c!} \left[ (\lambda)^{\alpha-1} \exp(-\beta\lambda) \frac{\beta^\alpha}{\Gamma(\alpha)} \right] d\lambda \\ &= \frac{\beta^\alpha t^c \Gamma(\alpha + c)}{c! \Gamma(\alpha) (\beta + t)^{\alpha+c}} \underbrace{\int_0^\infty \frac{(\beta + t)^{\alpha+c}}{\Gamma(\alpha + c)} \lambda^{\alpha+c-1} \exp(-(\beta + t)\lambda) d\lambda}_{=1} \\ &= \binom{\alpha + c - 1}{\alpha - 1} \left( \frac{t}{\beta + t} \right)^c \left( \frac{\beta}{\beta + t} \right)^\alpha, \end{aligned}$$

$$C(t) \sim NBin\left(\alpha, \frac{\beta}{\beta + t}\right)$$



## Expectation and Variance for Counts

Using the expressions of iterated expectation and variance  
(Held and Bové, 2014)

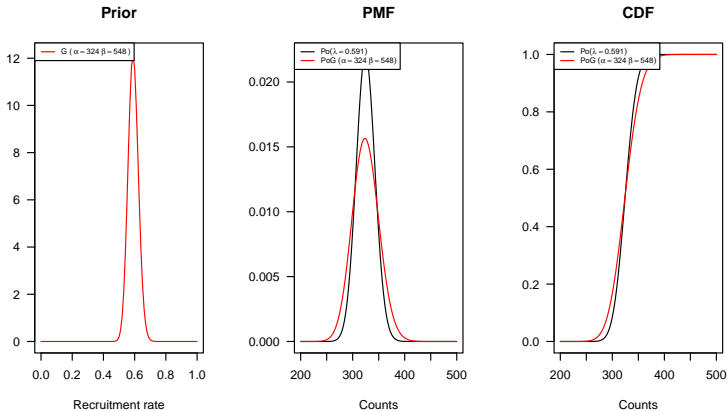
$$E(C(t)) = E_{\Lambda}[E_{C(t)}(C(t)|\Lambda)] = E_{\Lambda}[\Lambda t] = t\alpha/\beta$$

$$\begin{aligned} \text{Var}(C(t)) &= \text{Var}_{\Lambda}[E_{C(t)}(C(t)|\Lambda)] + E_{\Lambda}[\text{Var}_{C(t)}(C(t)|\Lambda)] \\ &= \text{Var}_{\Lambda}[\Lambda t] + E_{\Lambda}[\Lambda t] \\ &= t^2\alpha/\beta^2 + t\alpha/\beta = \frac{t\alpha(\beta + t)}{\beta^2} \end{aligned}$$

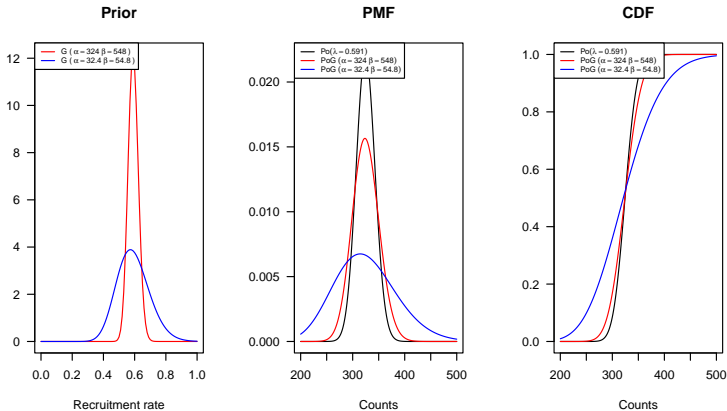


## Sensitivity Analysis

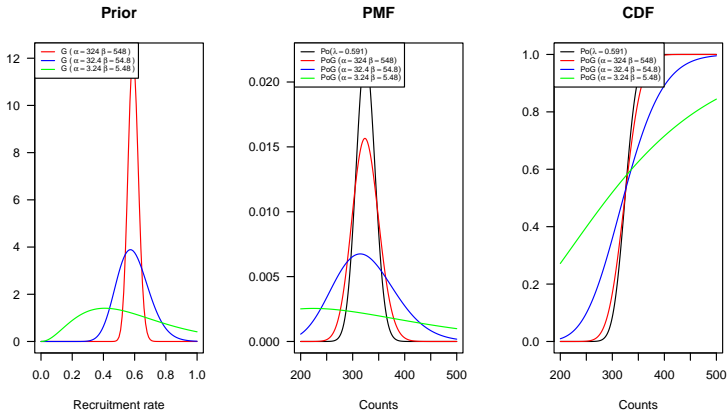
## Sensitivity Analysis



## Sensitivity Analysis



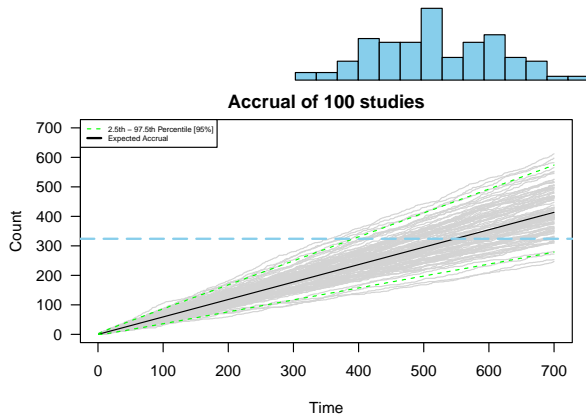
## Sensitivity Analysis





## Methods for Waiting Time

## Motivation Models for Waiting Time







## Models for Waiting Time until Target Sample Size $c$

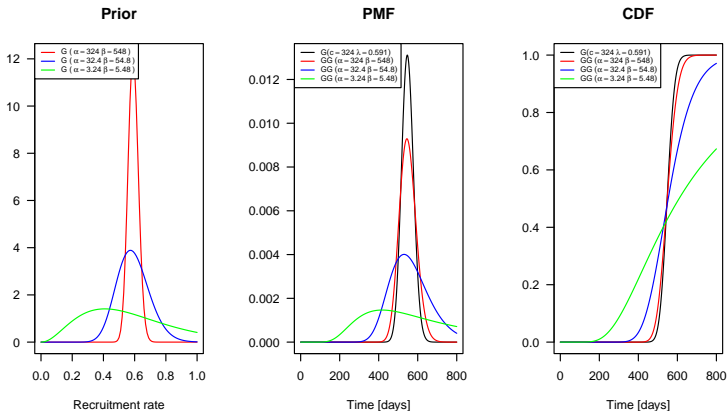
Methods	Time	Expectation	Variance	Aleatory	Epistemic
Expectation	$T(c) = c/\lambda$	$c/\lambda$	0	No	No
Erlang	$T(c) \sim G(c, \lambda)$	$c/\lambda$	$c/\lambda^2$	Yes	No
Gamma-Gamma	$T(c) \sim G(c, \Lambda); \Lambda \sim G(\alpha, \beta)$	$c \frac{\beta}{\alpha-1}$	$\frac{c\beta^2(c+\alpha-1)}{(\alpha-1)^2(\alpha-2)}$	Yes	Yes

Two versions of randomness of  $\Lambda$

Version 1  $\rightarrow$  GG distribution

Similar derivations as shown for counts

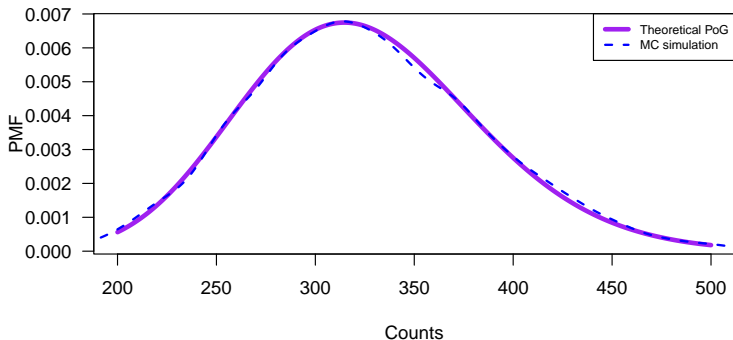
## Sensitivity Analysis for Waiting Time



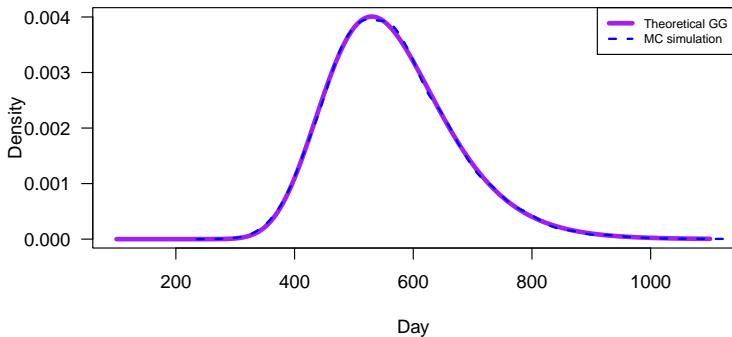


## Exact Methods vs MC simulations

## Exact Methods vs MC simulations – Poisson-Gamma Counts



## Exact Methods vs MC simulations – Gamma-Gamma Time





## Exact Methods vs MC simulations

Model	Estimated Probability	MCse	Exact Probability
$C(T) \sim \text{Po}(\lambda T)$	$P(C(T) \geq 324) = 0.5044$	0.005	0.5085
$C(T) \sim \text{PoG}(T, \alpha, \beta)$	$P(C(T) \geq 324) = 0.4799$	0.005	0.5008

Model	Estimated Probability	MCse	Exact Probability
$T(C) \sim G(C, \lambda)$	$P(T(C) \geq 548) = 0.4978$	0.005	0.4955
$T(C) \sim \text{GG}(C, \alpha, \beta)$	$P(T(C) \geq 548) = 0.5196$	0.005	0.5201

**Number of simulations:**  $M = 10^4$



## Aleatory VS Aleatory & Epistemic

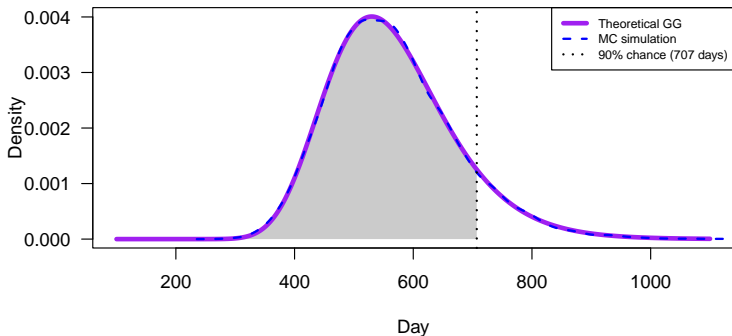
**90% chance** of accruing  $C_{target} = 324$  patients:

$M = 10^3$  from Carter's  $\rightarrow$  580 days (innacurate)

Erlang exact distribution  $\rightarrow$  588 days

Gamma-Gamma exact distribution  $\rightarrow$  707 days

## Aleatory & Epistemic







## Conclusions

- Visual tools
- Unified Notation
- Exact Methods
- Flexible Recruitment
- Practical Impact



## References

- Carter, R. E. (2004). Application of stochastic processes to participant recruitment in clinical trials. *Controlled Clinical Trials*, 25(5):429–436.
- Held, L. and Bové, D. S. (2014). *Applied Statistical Inference*. Springer.
- Piantadosi, S. (2024). *Clinical Trials: A Methodologic Perspective*. John Wiley & Sons.



**Thank you for your attention**