

# New perspectives on credible intervals and sample size determination

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# Overview

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- Notations
- Credible intervals
  - Equal-tails intervals (ETs)
  - Highest density intervals (HDs)
- Traditional SSD criteria
  - length criteria
  - location criteria
- New SSD criteria
  - A – Based on discrepancy between alternative exact intervals
  - B – Based on discrepancy between approximate and exact intervals

# Notations

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- parametric model

$$\{f_n(\cdot; \theta), \theta \in \Omega\}$$

- data

$$\mathbf{x}_n = (x_1, \dots, x_n)$$

- prior distribution

$$\pi(\theta)$$

- posterior distribution

$$\pi(\theta|\mathbf{x}_n) = \frac{f_n(\mathbf{x}_n|\theta)\pi(\theta)}{\int_{\Omega} f_n(\mathbf{x}_n|\theta)\pi(\theta)d\theta}$$

# Parameter

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$\theta$  represents

- unknown effect of a treatment
- unknown difference between effects of treatments
- ...

$\Theta$  is a **random variable**

## Credible interval

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A  $(1 - \gamma)$ -credible interval

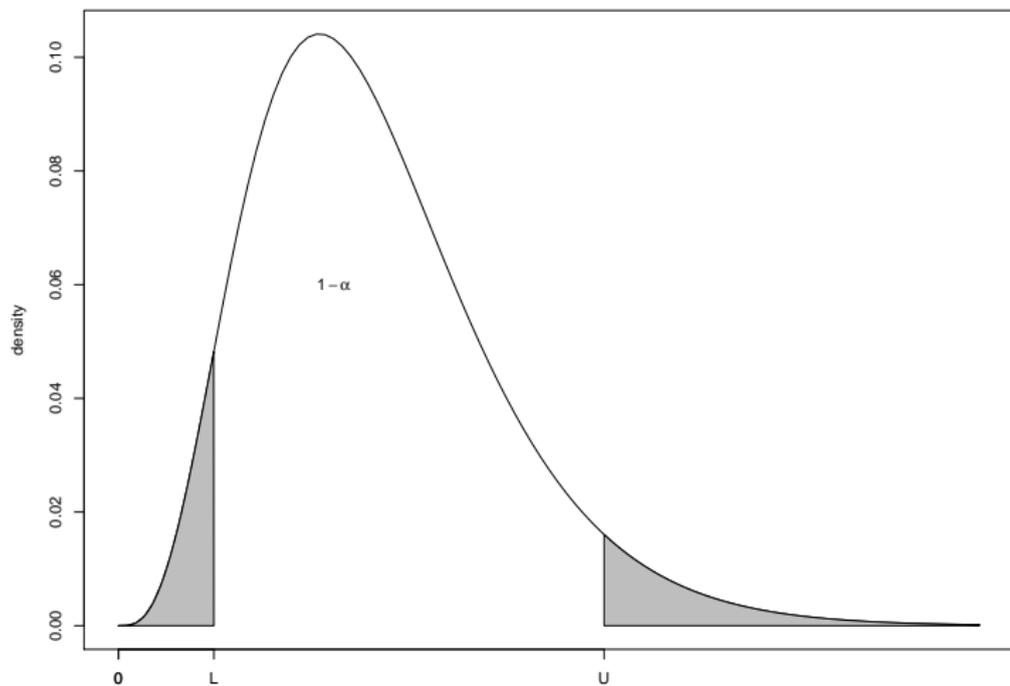
$$C(\mathbf{x}_n) = [\ell(\mathbf{x}_n), u(\mathbf{x}_n)]$$

- is subset of  $\Omega$  (parameter space)
- depends on  $\mathbf{x}_n$  and  $\pi(\theta)$
- has posterior probability equal to  $1 - \gamma$

$$\mathbb{P}[\Theta \in C|\mathbf{x}_n] = \int_C \pi(\theta|\mathbf{x}_n)d\theta = 1 - \gamma$$

# Credible interval

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# ET and HD credible intervals

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- Equal-tails intervals (ETs)

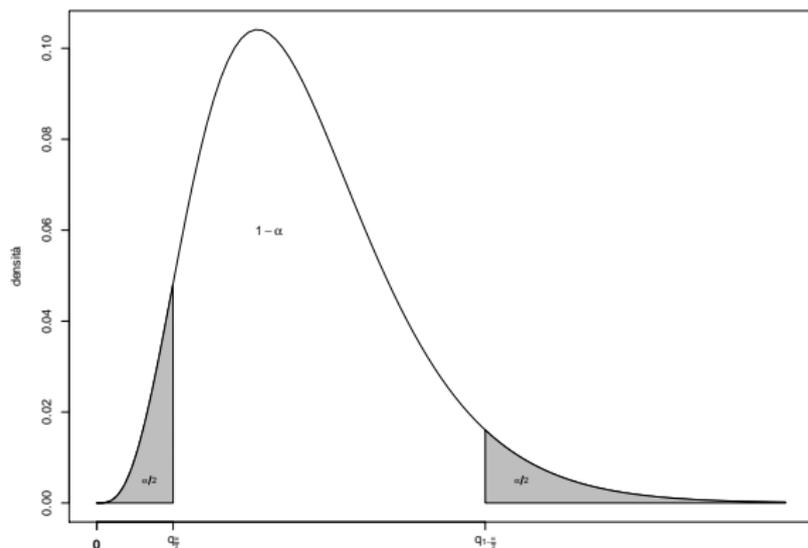
$$C_e(\mathbf{x}_n) = [\ell_e(\mathbf{x}_n), u_e(\mathbf{x}_n)]$$

- Highest (posterior) density intervals (HDs)

$$C_h(\mathbf{x}_n) = [\ell_h(\mathbf{x}_n), u_h(\mathbf{x}_n)]$$

## Equal-tails intervals

$$l_e = q_{\frac{\gamma}{2}} \quad u_e = q_{1-\frac{\gamma}{2}}$$



# Equal-tails intervals

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$$C_e = [l_e, u_e]$$

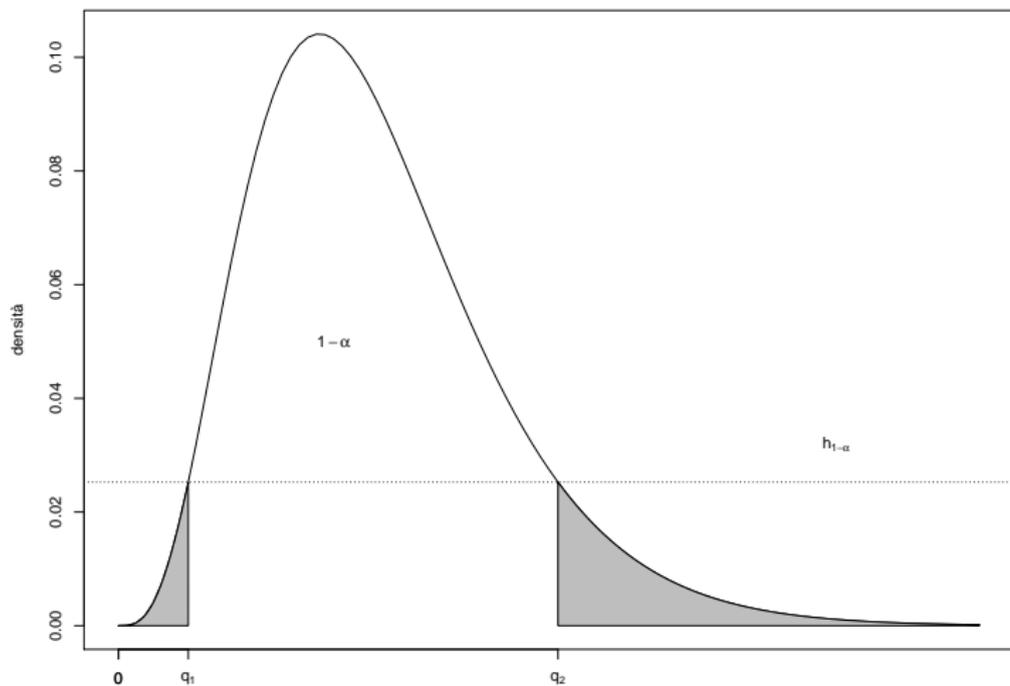
- $\mathbb{P}[l_e \leq \Theta \leq u_e | \mathbf{x}_n] = 1 - \gamma$
- $\mathbb{P}[\Theta \leq l_e | \mathbf{x}_n] = Pr[\Theta \geq u_e | \mathbf{x}_n] = \frac{\gamma}{2}$

## Features

- easy-to-obtain
- invariant under transformation
- non-optimal in length

# Highest posterior density intervals

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# Highest posterior density intervals

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$$C_h = [\ell_h, u_h]$$

- $\mathbb{P}[\ell_e \leq \Theta \leq u_e | \mathbf{x}_n] = 1 - \gamma$
- $\pi(\theta | \mathbf{x}_n) \geq h_{1-\gamma}, \quad \forall \theta \in C_h$

## Features

- not easy-to-obtain
- not invariant under transformation
- optimal in length

## Credible intervals and sample size

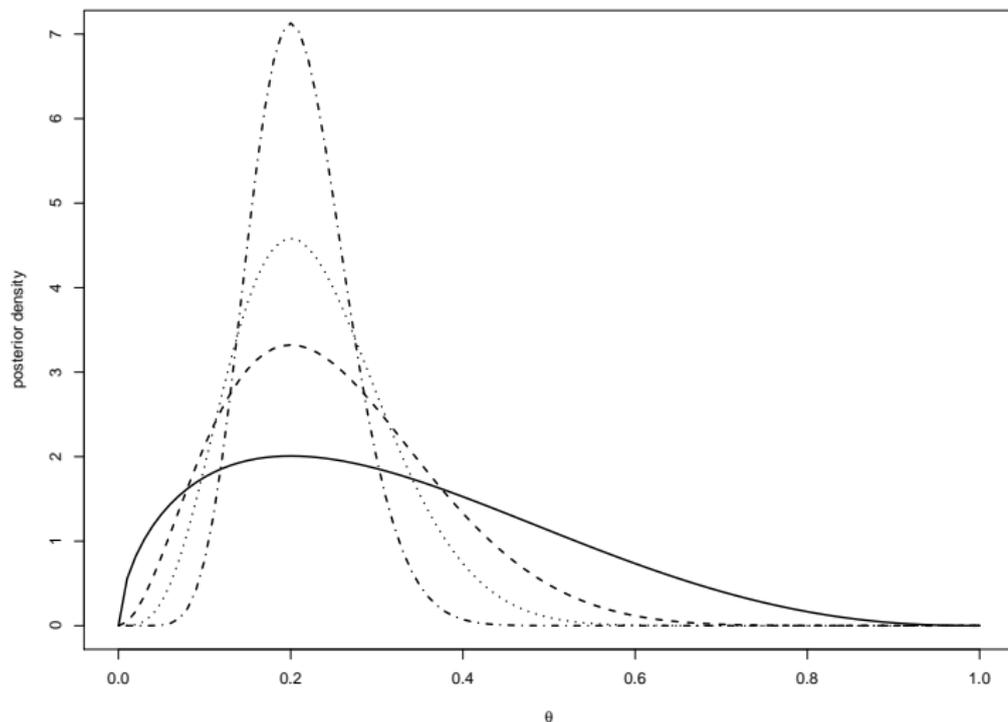
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As  $n$  increases, **posterior density** tends

- to concentrate around  $\theta^*$
- to symmetrize
- to normalize

# Credible intervals and sample size

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## Credible intervals and sample size

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As  $n$  increases, alternative credible intervals tend

- to become shorter and shorter
- to concentrate around  $\theta^*$  (true parameter value)
- to coincide

## Traditional Bayesian SSD criteria

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Consider

$$C(\mathbf{X}_n) = [\ell(\mathbf{X}_n), u(\mathbf{X}_n)]$$

where  $\mathbf{X}_n$  is random

Look at

- **Length**: want  $C$  to be short
- **Location**: want  $C$  either inside or outside a **range of equivalence**

## Criteria based on *Length*

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To control *precision* of intervals

Consider

$$C(\mathbf{X}_n) = [\ell(\mathbf{X}_n), u(\mathbf{X}_n)]$$

Define

$$\mathcal{L}(\mathbf{X}_n) = u(\mathbf{X}_n) - \ell(\mathbf{X}_n)$$

Determine

$$\min n \in \mathbb{N} : \mathbb{E}[\mathcal{L}(\mathbf{X}_n)] < \epsilon$$

## Criteria based on *Location* (cont.)

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To control the **location** of  $C$  w.r.t. a **range of equivalence**

$$\mathcal{I} = [\theta_i, \theta_s] \subset \Omega$$

- Superiority trials

$$\min n \in \mathbb{N} : \mathbb{E}[\ell(\mathbf{X}_n)] > \theta_s$$

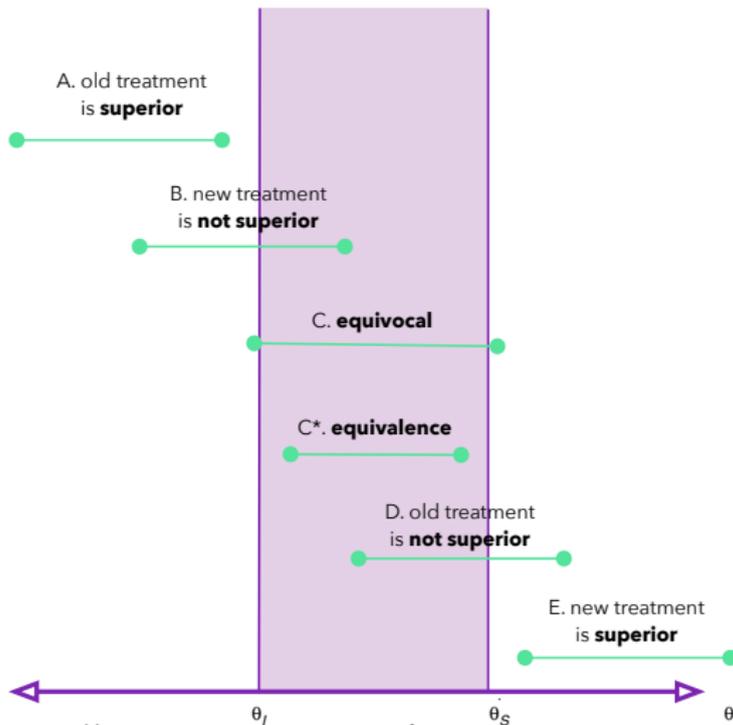
- Inferiority trials

$$\min n \in \mathbb{N} : \mathbb{E}[u(\mathbf{X}_n)] < \theta_i$$

- Equivalence trials

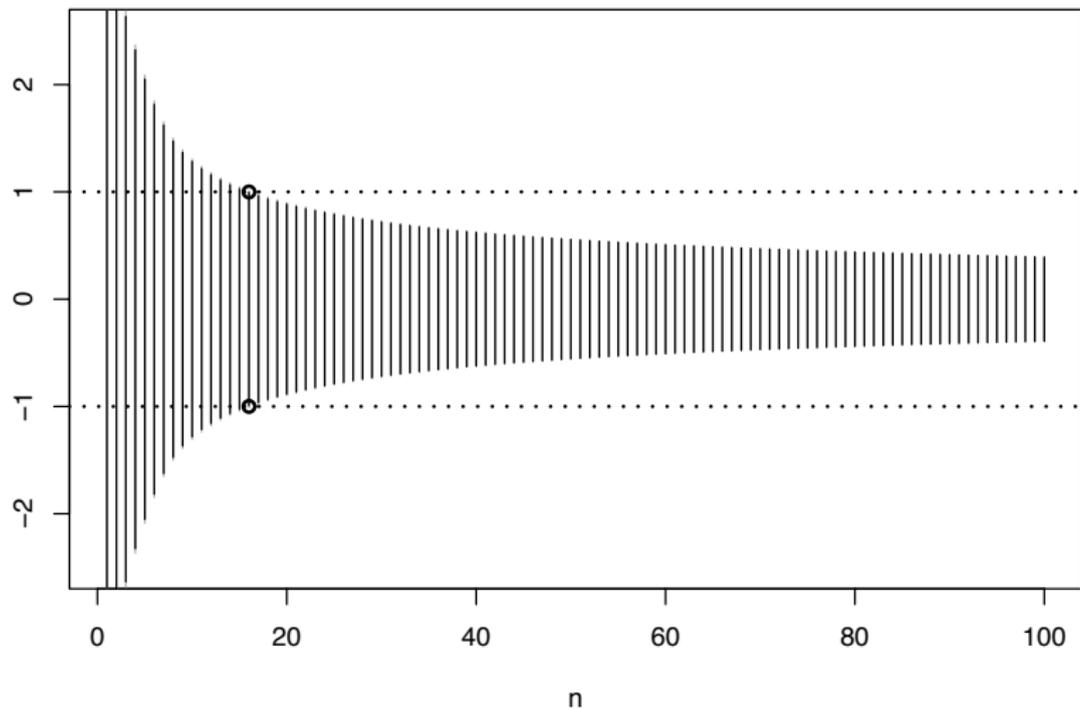
$$\min n \in \mathbb{N} : \mathbb{E}[\ell(\mathbf{X}_n)] > \theta_s \quad \text{and} \quad \mathbb{E}[u(\mathbf{X}_n)] < \theta_i$$

## Criteria based on *Location*



## Example (equivalence)

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# Alternative approaches

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Based on

consensus

A) between alternative **exact** intervals

- equal-tails intervals
- highest density intervals

B) between **approximate** and **exact** intervals

- intervals based on normal approximations
- highest density intervals and/ or equal-tails intervals

# Motivations

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Why do we want **consensus**?

- A) want an **exact** interval with the **good properties** of both ETs and HDs
  
- B) want an **approximate** interval that is
  - easy to compute
  - close to an exact credible interval

## Part A – Consensus between ETs and HDs

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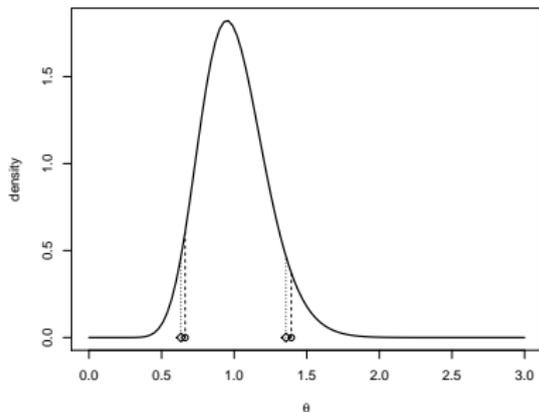
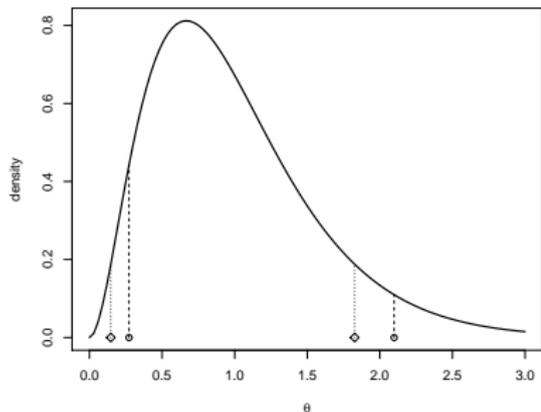
Ideal

an interval with good properties of both ETs and HDs

## Bad news - good news!

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- **small  $n$** : ETs and HDs do not coincide (in general)
- **large  $n$** : closer and closer



## Idea!

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proceed as follows

- quantify the **discrepancy** between  $C_e$  and  $C_h$
- consider the discrepancy as a **random variable**
- evaluate the **expected value** of the discrepancy
- select the **smallest sample size** such that the expected discrepancy is small

## Discrepancy between $C_e$ and $C_h$

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more formally

- $D_n(\mathbf{x}_n) = d(C_e, C_h)$
- $D_n(\mathbf{X}_n)$  as a random variable
- $e_n^D = \mathbb{E}[D_n]$
- $n^*$ , optimal sample size as

$$n^* = \min\{n \in \mathbb{N} : e_n^D \leq \epsilon\}$$

# Types of discrepancies between intervals

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Alternative measures  $D_n$ , based on

- distance between **bounds** of  $C_e$  and  $C_h$
- difference in **probabilities of tails** between  $C_e$  and  $C_h$

## “Bounds” discrepancy

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- $B_n = |\ell_e - \ell_h| + |u_e - u_h|$
- $e_n^B = \mathbb{E}[B_n]$
- $n^* = \{\min \mathbb{N} : e_n^B < \epsilon\}$

BUT:  $e_n^B$  is not a relative measure

## “Tails” discrepancy

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### Idea

- note that as  $n$  increases  $C_h$  becomes an ET interval
- define a measure  $T_n$  such that is
  - equal to zero if  $C_h$  has tails-probability both equal to  $\frac{\gamma}{2}$
  - greater than zero if the tails-probability of  $C_h$  are not equal to  $\frac{\gamma}{2}$

## “Tails” discrepancy

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It can be easily shown that

$$T_n(\mathbf{x}_n) = \frac{|2F_n(\ell_h|\mathbf{x}_n) - \gamma|}{\gamma}$$

is a [relative measure of discrepancy](#) between  $C_h$  and the ET interval (where  $F_n(\cdot|\mathbf{x}_n)$  is the posterior c.d.f)

## “Tails” discrepancy and SSD

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Therefore

$$n^* = \{\min \mathbb{N} : e_n^T < \epsilon\}$$

where

$$e_n^T = \mathbb{E}[T_n]$$

## Main comments

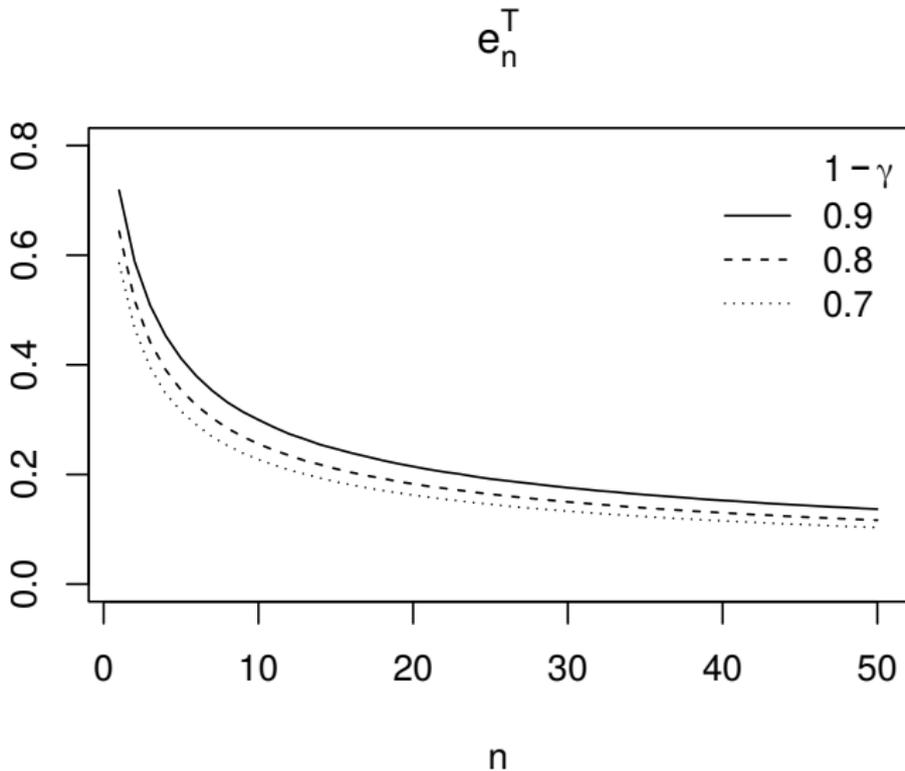
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$n^*$  increases with

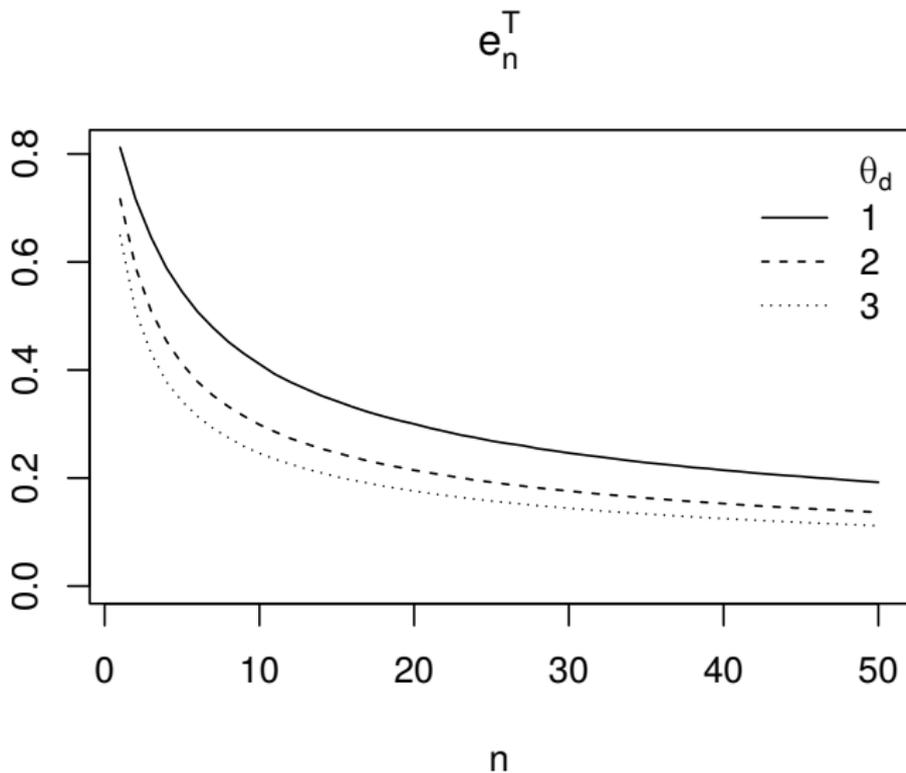
- credibility level  $1 - \gamma$
- asymmetry of the likelihood function
- asymmetry of the prior density

## Example (Poisson model): effect of $1 - \gamma$

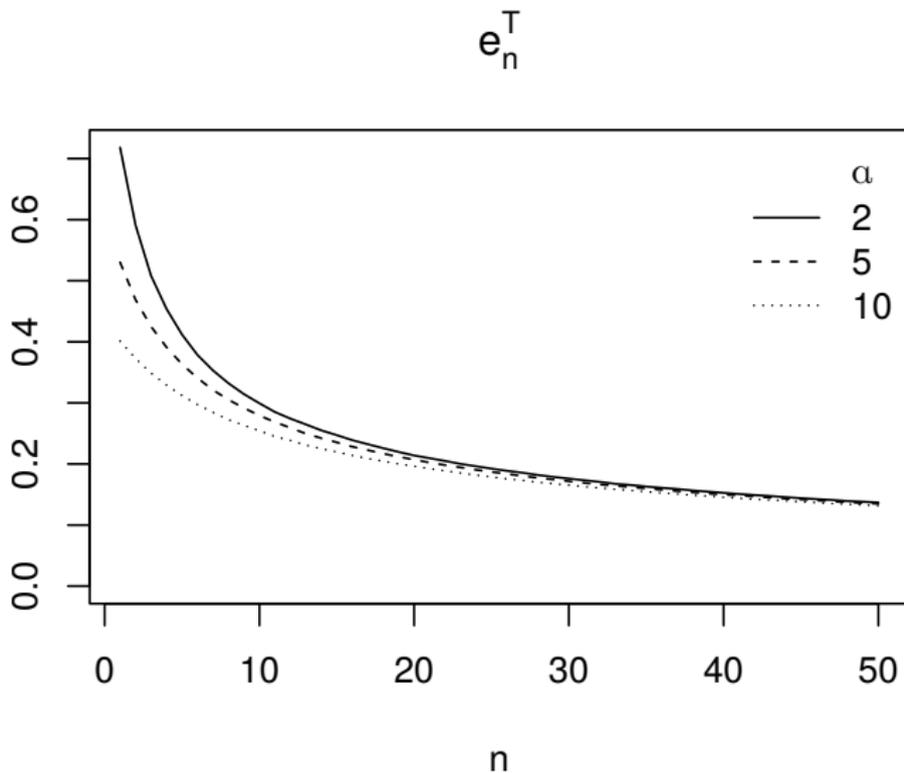
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## Example (Poisson model): effect of likelihood



## Example (Poisson model): effect of prior



## Example (Poisson model): relationships with skewness

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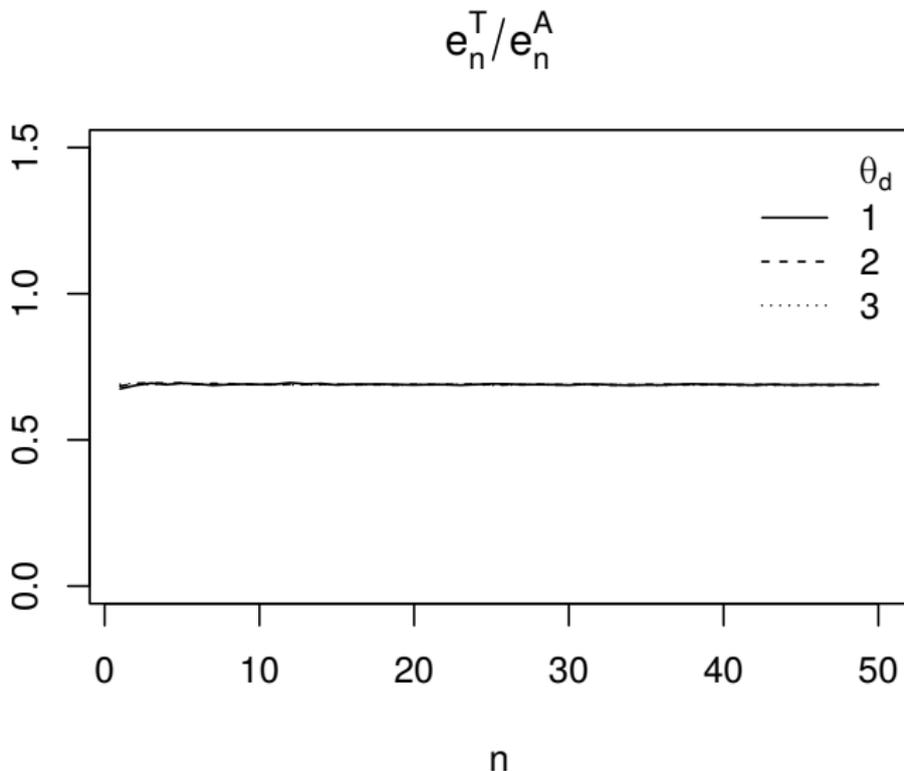
- $T_n(\mathbf{x})$  is a measure of skewness of  $\pi(\theta|\mathbf{x}_n)$
- Equivalent to

$$A_n(\mathbf{x}_n) = \frac{\bar{\mu}_3(\mathbf{x}_n)}{\sigma^3(\mathbf{x}_n)}$$

- Consequence

$$\frac{e_n^T}{e_n^A} \simeq 1$$

## Example (Poisson model): relationships with skewness



## Example (Poisson model): optimal sample sizes

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$n^*$  using  $e_n^T$

	$1 - \gamma$	0.9			0.8			0.7		
	$\epsilon$	0.15	0.2	0.25	0.15	0.2	0.25	0.15	0.2	0.25
$\theta_d$	1	83	47	30	60	34	22	48	27	17
	2	42	24	15	30	17	11	24	14	9
	3	28	16	10	21	12	8	16	9	6

$\theta_d =$  design value of  $\theta$

## Part B – Approximate and exact intervals

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Let

$$\tilde{C} = [\tilde{\ell}_n, \tilde{u}_n]$$

be an **approximate credible interval**.

For instance the **likelihood interval**

$$\tilde{C} = [\hat{\theta} - z_{1-\frac{\alpha}{2}} \sqrt{I_n(\hat{\theta})^{-1}}, \hat{\theta} + z_{1-\frac{\alpha}{2}} \sqrt{I_n(\hat{\theta})^{-1}}]$$

## Approximate intervals

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Want minimum  $n$  such that  $\tilde{C}$  is approximately

- $(1 - \gamma)$ -credible interval
- ET interval
- HD interval

## Measures of discrepancy for approximate intervals

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Define 3 measures of discrepancy

$$P_n, T_n, B_n$$

such that

- small  $P_n \Rightarrow \tilde{C}$  is approximately a  $(1 - \gamma)$  CI
- small  $T_n \Rightarrow \tilde{C}$  is is approximately ET
- small  $B_n \Rightarrow \tilde{C}$  is is approximately HD

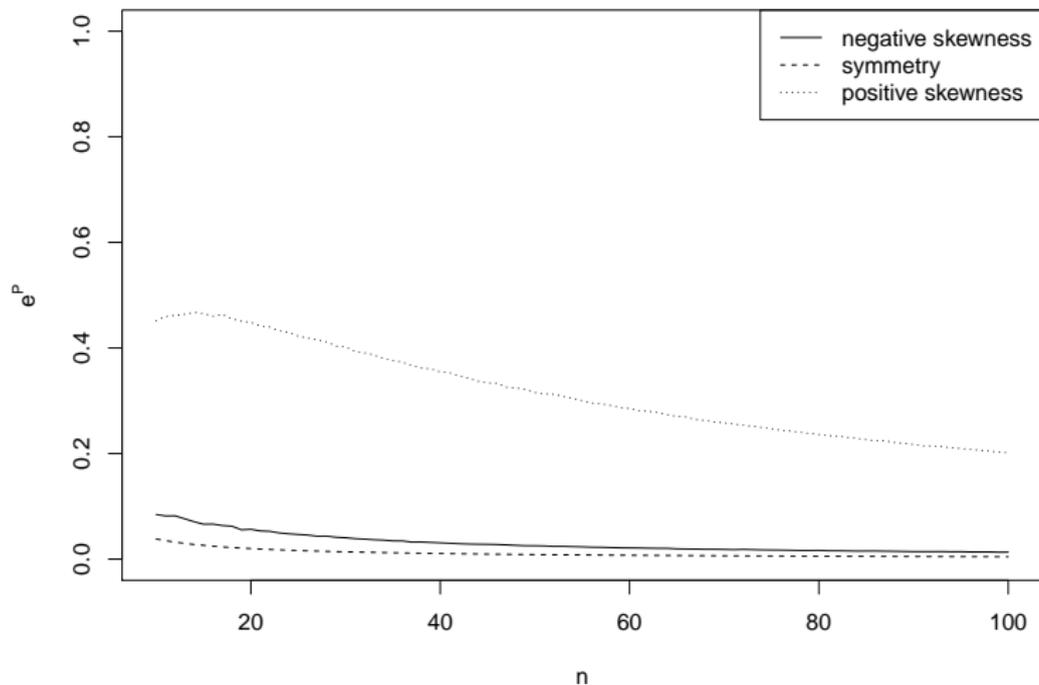
## SSD for approximate intervals

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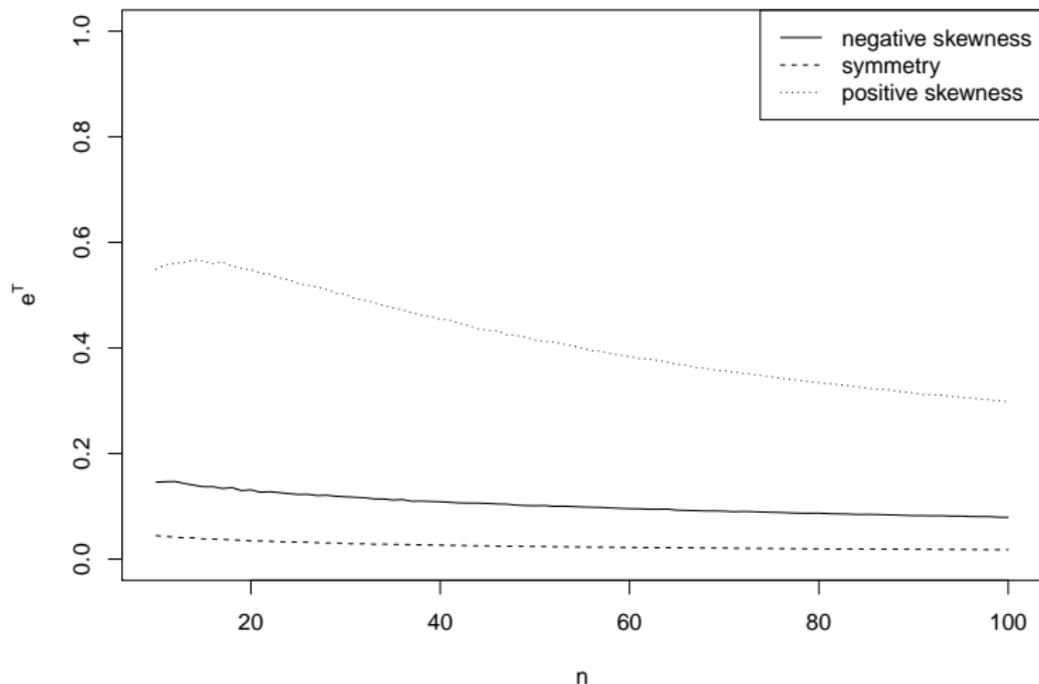
For SSD, consider

- $e_n^P = \mathbb{E}[P_n]$
- $e_n^T = \mathbb{E}[T_n]$
- $e_n^B = \mathbb{E}[B_n]$

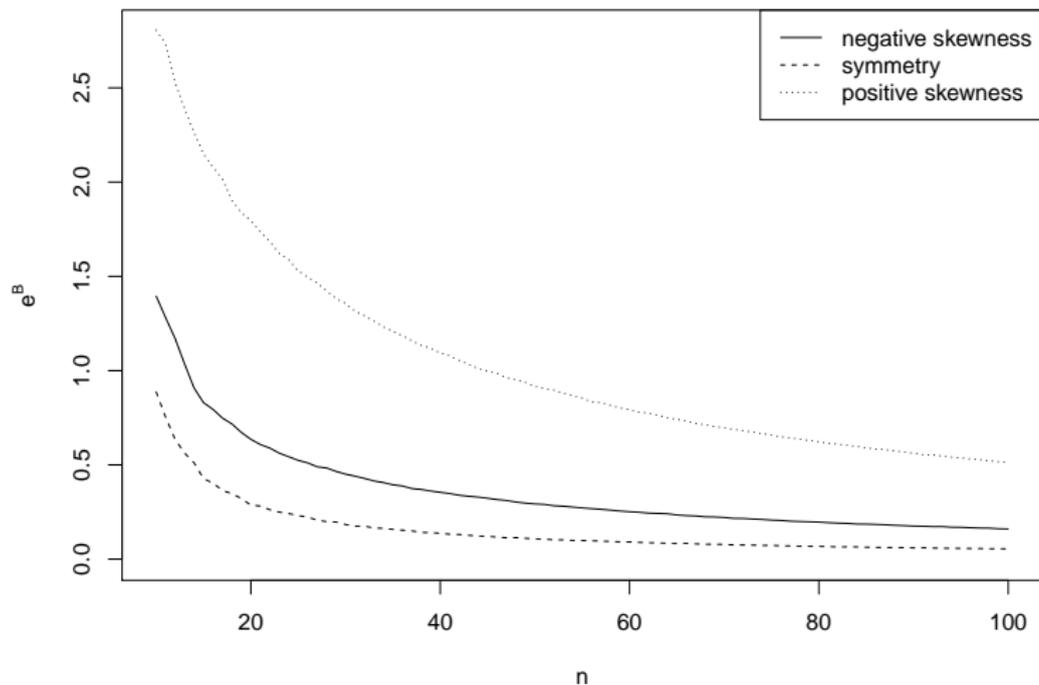
## Example: approx CI for log-odds ( $e_n^P = \mathbb{E}[P_n]$ )



## Example: approx CI for log-odds ( $e_n^T = \mathbb{E}[T_n]$ )

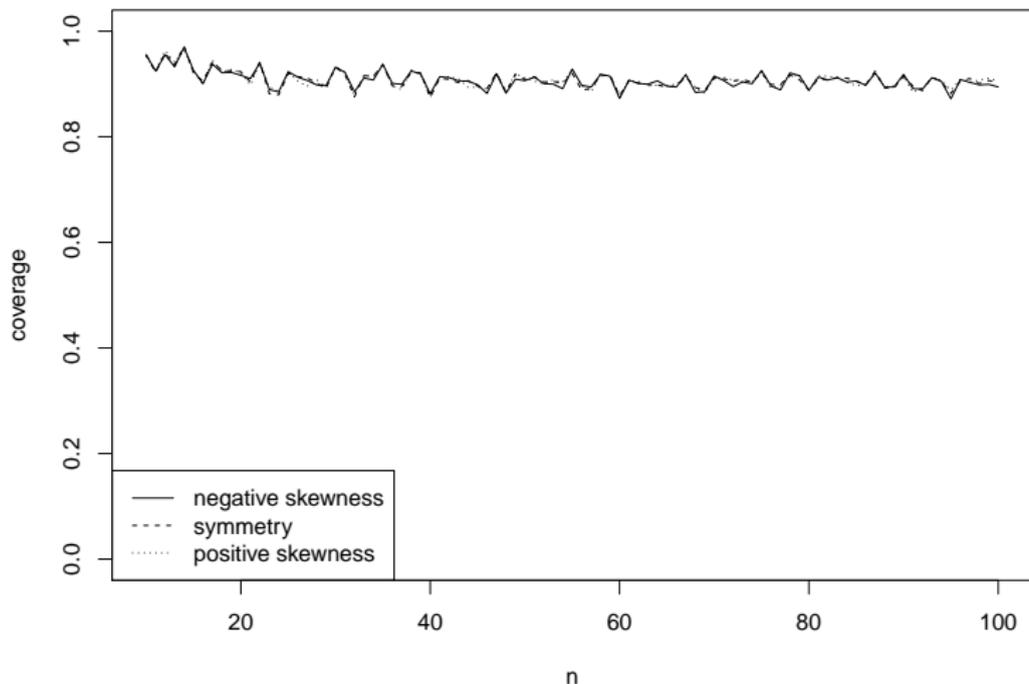


## Example: approx CI for log-odds ( $e_n^B = \mathbb{E}[B_n]$ )



## Example: approx CI for log-odds (coverage)

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## Final comments

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- 1 Selection of measures of discrepancy
- 2 Difficulties in fixing some thresholds ( $B_n$  is not a relative measure)
- 3 Relationships with other SSD criteria
- 4 Multiparametric problems

## References: general

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Cao J., Lee J. J. and Albert, S. (2009). Comparison of Bayesian sample size determination criteria: ACC, ALC, and WOC. *Journal of Statistical Planning and Inference*, 139, 4111 - 4122.

Joseph L. and Wolfson D. (1997). Interval-based versus decision theoretic criteria for the choice of sample size. *Journal of the Royal Statistical Society. Series D (The Statistician)*, 46, 145 - 149.

## References: range of equivalence

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Brutti P. and De Santis F. (2008). Robust Bayesian sample size determination for avoiding the range of equivalence in clinical trials. *Journal of Statistical Planning and Inference*, 138, 1577 - 1591.

Gubbiotti S. and De Santis, F. (2011). A Bayesian method for the choice of the sample size in equivalence trials. *Australian and New Zealand Journal of Statistics*, 53, 443 - 460.

## References: control of discrepancy

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De Santis F. and Gubbiotti S. (2019). Progressive overlap of equal-tails and highest posterior density intervals. Submitted.

De Santis F. and Gubbiotti S. (2019). Sample size determination for calibrated approximate credible intervals in clinical trials. In progress.

## “Tails” discrepancy

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Idea: check when the HD interval becomes an ET interval

- $C_h = [\ell_h, u_h]$
- $F_n(\cdot | \mathbf{x}_n)$  (posterior c.d.f)
- $F_n(\ell_h | \mathbf{x}_n)$       and       $1 - F_n(u_h | \mathbf{x}_n)$
- $T_1 = |F_n(\ell_h | \mathbf{x}_n) - \gamma/2|$       and       $T_2 = |1 - F_n(u_h | \mathbf{x}_n) - \gamma/2|$

## “Tails” discrepancy (cont.)

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Note that

- $T_1 + T_2 = |2F_n(\ell_e|\mathbf{x}_n) - \gamma|$
- $0 \leq |2F_n(\ell_e|\mathbf{x}_n) - \gamma| \leq \gamma$

Therefore a **relative measure of discrepancy** is

$$T_n(\mathbf{x}_n) = \frac{|2F_n(\ell_h|\mathbf{x}_n) - \gamma|}{\gamma}$$

## “Tails” discrepancy (cont.)

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- $T_n(\mathbf{x}_n) = \frac{|2F_n(\ell_h|\mathbf{x}_n) - \gamma|}{\gamma}$
- $e_n^T = \mathbb{E}[T_n]$
- $n^* = \{\min \mathbb{N} : e_n^T < \epsilon\}$

## Measures of discrepancy for approximate intervals

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$\tilde{C}(\mathbf{x}_n) = [\tilde{\ell}_n, \tilde{u}_n]$  is

approximately  $1 - \gamma$  if

$$P_n = |F_n(\tilde{u}_n) - F_n(\tilde{\ell}_n) - (1 - \gamma)| \quad \text{small}$$

approximately ET if

$$T_n = |F_n(\tilde{u}_n) - \frac{\gamma}{2}| + |1 - F_n(\tilde{\ell}_n) - \frac{\gamma}{2}| \quad \text{small}$$

approximately HD if

$$B_n = |\tilde{\ell}_n - \ell_h| + |\tilde{u}_n - u_h|$$