

Motivation

- ▶ Post-selection inference (PoSI) is well established for linear regression, where it provides valid inference after arbitrary model selection by controlling coverage over a prespecified class of submodels.
- ▶ This perspective can be transferred to Cox regression, but its practical implications in the survival setting are not yet well understood.
- ▶ We study two Cox-PoSI constructions:
 - standardized max- $|Z|$,
 - score-based supremum.
- ▶ Main finding: both procedures are asymptotically valid, but both necessarily produce larger critical values than full-model simultaneous inference.

Cox-PoSI framework

Setup

- ▶ Survival data $(T_i, \Delta_i, X_i)_{i=1}^n$ with independent censoring.
- ▶ Candidate submodels

$$m \subseteq \{1, \dots, p\}, \quad m \in \mathcal{M}_k.$$

- ▶ For each submodel m , let $\hat{\beta}_m$ denote the Cox partial likelihood estimator.
- ▶ Global index set

$$\mathcal{I} = \{(m, j) : m \in \mathcal{M}_k, j \in m\}.$$

Two Cox-PoSI statistics

- ▶ Standardized version:

$$T_n^Z = \max_{(m,j) \in \mathcal{I}} |Z_{m,j,n}|.$$

- ▶ Score-based version:

$$T_n^S = \max_{(m,j) \in \mathcal{I}} |S_{m,j,n}|.$$

- ▶ Both converge to maxima of centered Gaussian limits.
- ▶ Hence simultaneous confidence statements remain valid after arbitrary selection within \mathcal{M}_k .

Structural inflation

- ▶ If the full model is contained in \mathcal{M}_k , then its coordinates are part of the PoSI index set.
- ▶ Therefore,

$$T_n^Z \geq T_{\text{full}}^Z, \quad T_n^S \geq T_{\text{full}}^S,$$

and thus

$$K_{1-\alpha}^Z \geq K_{1-\alpha, \text{full}}^Z, \quad K_{1-\alpha}^S \geq K_{1-\alpha, \text{full}}^S.$$

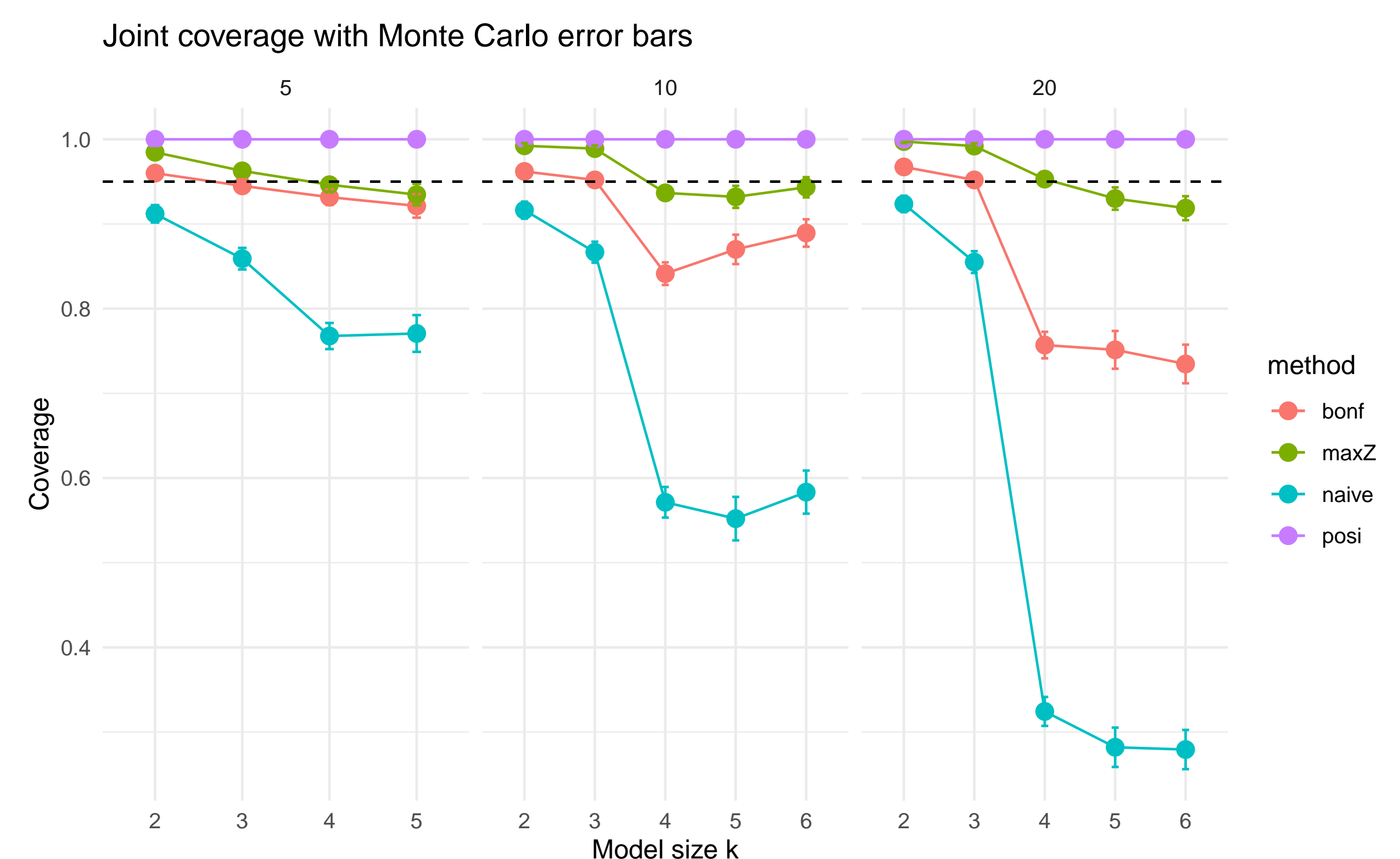
- ▶ Inflation is structural:
 - not a numerical artifact,
 - not a tuning issue,
 - caused by maximizing over a larger index set.
- ▶ Empirically, the growth is close to $\sqrt{2 \log |\mathcal{I}|}$.

Key message:

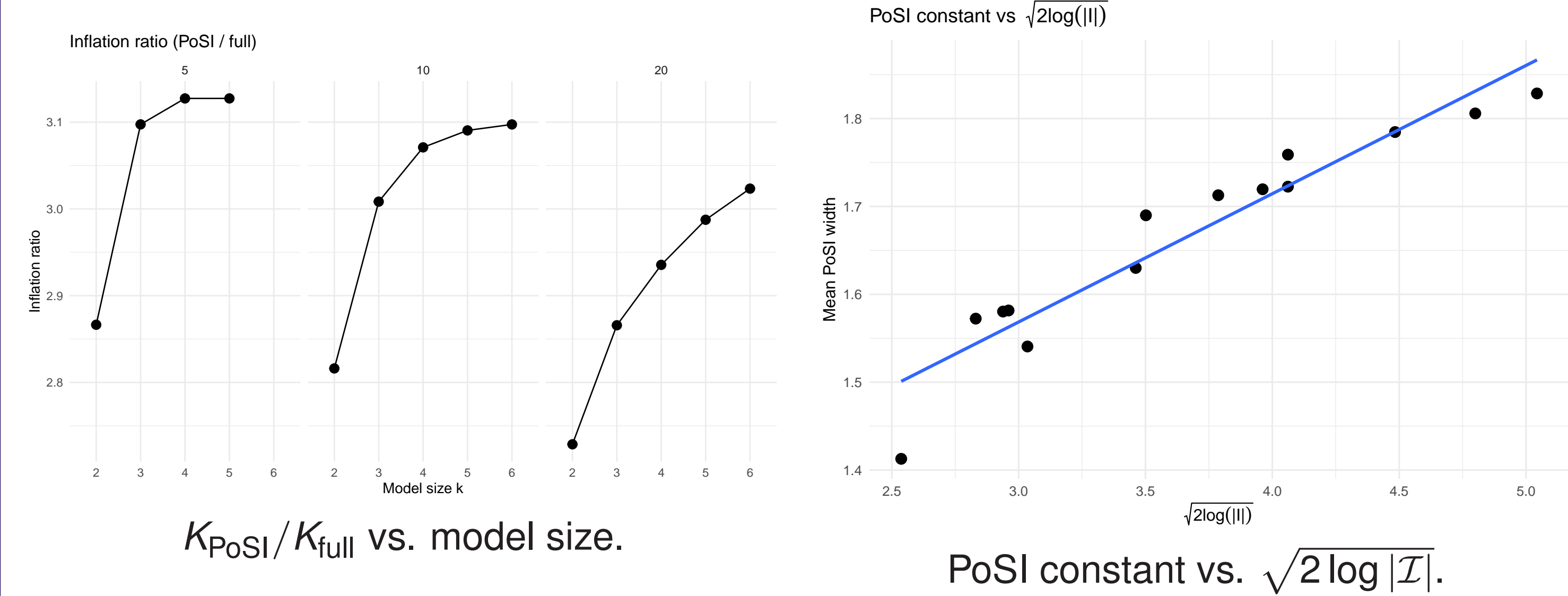
PoSI controls inference over the whole candidate model class, not only over the selected model.

Simulation study

- ▶ Right-censored data with $n = 200$, $p = 10$, three non-zero coefficients, and independent censoring.
- ▶ Candidate model class \mathcal{M}_k : all submodels up to size k .
- ▶ Naive post-selection intervals under-cover and deteriorate as k increases.
- ▶ Cox-PoSI maintains joint coverage, but at the price of substantially larger critical values.



Joint coverage over \mathcal{M}_k .



Real data example: TCGA-BRCA

Data and selected model

- ▶ Clinical survival data from the TCGA Breast Invasive Carcinoma cohort.
- ▶ Candidate covariates: age and stage-derived indicators.
- ▶ Candidate model class: all submodels up to size $k = 3$.
- ▶ AIC selected a model with age, age², and stage IV.

Interpretation

- ▶ Naive and full-model simultaneous intervals remain relatively informative.
- ▶ PoSI intervals are much wider.
- ▶ The application confirms the structural cost already seen in simulation.

Critical constants

Method	Constant
Naive	1.96
Full max- $ Z $	2.84
PoSI	11.44

Stage IV hazard ratio interval

Method	95% interval
Naive	[6.05, 19.31]
Full max- $ Z $	[4.65, 25.09]
PoSI	[0.36, 319.94]

Main practical message

PoSI is valid, but expensive.

Even for a moderate model class, the PoSI critical constant is much larger than the naive and full-model alternatives. This produces intervals that are theoretically valid but may become practically too wide to be useful.

For stage IV, the PoSI interval expands to

$$[0.36, 319.94],$$

illustrating the loss of informativeness very clearly.

- ▶ Moderate model class
- ▶ Strong inflation of critical constants
- ▶ Strong widening of intervals
- ▶ Same qualitative pattern as in simulation

Take-home messages

- ▶ Cox-PoSI yields valid simultaneous inference over finite model classes.
- ▶ This gives valid post-selection inference after arbitrary selection within \mathcal{M}_k .
- ▶ The price is unavoidable structural inflation of critical values.
- ▶ In practice, Cox-PoSI is best viewed as a benchmark for post-selection uncertainty rather than a routine default.

References

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