Bayesian parameter synthesis for markov population model.

Nhat-Huy Phung

University of Konstanz

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Universität Konstanz





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Introduction.

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Preliminaries

Katoen [1] definition

Definition (Discrete-time Markov Chain)

Katoen [1] definition

Definition (PCTL)

Model checking PCTL

Katoen [1] definition

Definition (PCTL)

Model checking PCTL

Katoen [1] definition

Definition (PCTL)

Katoen [1] definition

Definition (Parametric Discrete-time Markov Chain)

Katoen [1] definition

Definition (Parameter synthesis)

Given a finite-state parametric Markov model, find the parameter values for which a given reachability property exceeds (or is below) a given threshold β .

Parameter synthesis and Parameter inference

What are the differences between parameter inference and parameter synthesis?

	Input	Ouput
Parameter	Model $\mathcal{M}_{ heta}$	Parameter estimation
inference	Observed data D_{obs}	$\hat{ heta}$
Parameter	Model $\mathcal{M}_{ heta}$	$(\theta_1,\ldots,\theta_N)$
synthesis	Reachability property Φ	$\forall heta_i \in (heta_1, \dots, heta_N) : \mathcal{M}_{ heta_i} \models \Phi$

Synthesis: input: model \mathcal{M}_{θ} and property Φ , output is the set of satisfying parameters θ Inference: input: model \mathcal{M}_{θ} and observed data D_{obs} , output is the set of parameter estimation $\hat{\theta}$ This thesis: combines parameter inference and synthesis into data-informed, Bayesian parameter synthesis frameworks.

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Frameworks

Model checking step

A property Φ is

- a bounded reachability property and
- specified by PCTL.

Checking a model \mathcal{M}_{θ} against Φ

- Rational fucntion evaluation
- Statistical model checking

Parameter synthesis step

Monte-Carlo Markov chain framework, basically modified Metropolis-Hasting algorithm.

Algorithm 1 Markov Chain Monte-Carlo with rational functions

```
procedure MCMC-RF
           Init \theta from prior distribution \pi(\theta).
 2:
 3.
           i \leftarrow 1
 4:
           while i \le N do
                 Draw \theta_{cand} from Q(\theta|\theta_1,\ldots,\theta_i)
 5:
                 if P(D_{obs}|\theta_{cand}) > P(D_{obs}|\theta_i) then
 6:
 7:
                      Accept \theta_{cand} if \mathcal{M}_{\theta_{cand}} \models \Phi
                      i \leftarrow i + 1
 8:
                end if
 g٠
10:
           end while
11:
           Return (\theta_1, \ldots, \theta_N)
      end procedure
```

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Parameter synthesis step

Algorithm 2 Sequential Monte-Carlo with rational functions

```
procedure SMC-RF
 2:
            Init \theta_1, \ldots, \theta_N from prior distribution \pi(\theta).
 3:
            Set w_i \leftarrow P(D_{obs}|\theta_i), 1 < i < N
 4:
            for t \in (1, \ldots, M) do
 5:
                 Normalize w_1^t, \ldots, w_N^t
                 Sample with replacement \theta_1^t, \dots, \theta_N^t from \theta_1^{t-1}, \dots, \theta_N^{t-1}
 6:
 7:
                 for \theta_i \in (\theta_1^t, \dots, \theta_N^t) do
 8:
                       \theta_i \leftarrow MCMC - RF(\theta_i), \ Q = F_i(\theta | \theta_1^{t-1}, \dots, \theta_N^{t-1})
 9.
                 end for
10:
            end for
11:
            Return (\theta_1, \ldots, \theta_N)
12: end procedure
```

Parameter synthesis step

Algorithm 3 Sequential Monte-Carlo with simulations

```
1: procedure RF-SMC
 2:
            Init \theta_1, \ldots, \theta_N from prior distribution \pi(\theta).
 3:
            for t \in (1, \ldots, M) do
                  Sample with replacement \theta_1^t, \dots, \theta_N^t from \theta_1^{t-1}, \dots, \theta_N^{t-1}
 4:
 5:
                  for \theta_i \in (\theta_1^t, \dots, \theta_N^t) do
                       Draw \theta_{cand} from F_i(\theta|\theta_1^{t-1},\ldots,\theta_N^{t-1})
 6:
 7:
                       if Statistical Model Checking \mathcal{M}_{\theta_{cand}} \models \Phi then
 8:
                             Simulate D_{sim} from \mathcal{M}_{\theta_{cond}}
                             if Distance(D_{sim}, D_{obs} < \epsilon) then
 9.
10:
                                  Update \theta_i \leftarrow \theta_{cand}
11:
                             end if
12:
                       end if
13:
                  end for
14:
            end for
15:
            Return (\theta_1, \ldots, \theta_N)
```

References I

[1] Joost-Pieter Katoen. "The probabilistic model checking landscape". In: *Proceedings of the 31st Annual ACM/IEEE Symposium on Logic in Computer Science.* 2016, pp. 31–45.