

ABSTRACT We are interested in characterising the class $[\mathcal{T}, \Theta_{\mathcal{T}}]$ of *statistically equivalent* representations of a probability tree model $\mathbb{P}_{(\mathcal{T}, \Theta_{\mathcal{T}})}$. This problem is analogous to the one of finding one *essential graph* D^* which indexes a class $[D]$ of DAG representations for a BN model \mathbb{P}_D . In the case of staged trees (and chain event graphs), no such graphical result is available. We thus present an alternative algebraic characterisation, showing that an *interpolating polynomial* $c_{\mathcal{T}} \in \mathbb{R}[\Theta_{\mathcal{T}}]$ uniquely identifies $[\mathcal{T}, \Theta_{\mathcal{T}}]$. This is important for computational reasons, in model selection and for causal discovery.

THE PROBABILITY TREE

Let the graph $\mathcal{T} = (V, E)$ be an *event tree* and $\Theta_{\mathcal{T}} = \{\theta_v \mid v \in V\}$ a set of parameter vectors $\theta_v = (\theta(e) \mid e \in E(v)) \in \Delta_{\#E(v)-1}^{\circ}$. Then,

$$\pi_{\theta}(\lambda) = \prod_{e \in E(\lambda)} \theta(e)$$

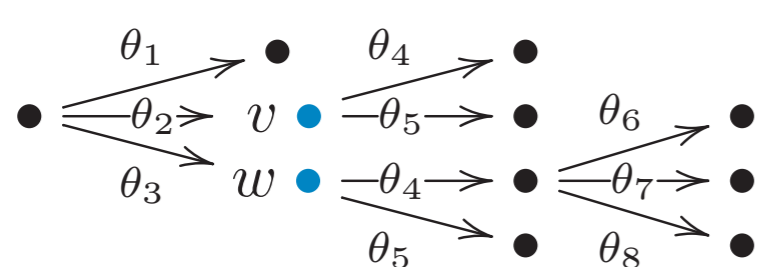
defines a strictly positive probability mass function on the root-to-leaf paths $\lambda \in \Lambda(\mathcal{T})$. We call the pair $(\mathcal{T}, \Theta_{\mathcal{T}})$ a *probability tree*.

Such a labelled graph is a picture for a discrete parametric statistical model

$$\mathbb{P}_{(\mathcal{T}, \Theta_{\mathcal{T}})} = \left\{ \pi_{\theta} \mid \theta \in \Delta_{d(E)}^{\circ} \right\} \subseteq \Delta_{\#\Lambda(\mathcal{T})-1}^{\circ}$$

Denote by $[\mathcal{T}, \Theta_{\mathcal{T}}]$ the set of possible representations of $\mathbb{P}_{(\mathcal{T}, \Theta_{\mathcal{T}})}$. We say that two trees $(\mathcal{T}, \Theta_{\mathcal{T}}), (\mathcal{S}, \Theta_{\mathcal{S}}) \in [\mathcal{T}, \Theta_{\mathcal{T}}]$ are *statistically equivalent*.

Two vertices $v, w \in V$ are in the same *stage* if and only if their parameter vectors coincide, $\theta_v = \theta_w$. We then write $v \sim w$.



Think of this as a sort of *conditional independence* statement, as formalised in Thwaites & Smith (2015).

The stage structure of a tree $(\mathcal{T}, \Theta_{\mathcal{T}})$ may be captured in the *stage ideal*

$$I_{\mathcal{T}} = \sum_{v \sim w} \langle \theta_{v,i} - \theta_{w,i} \mid i = 1, \dots, \#E(v_i) \rangle$$

in a ring $\mathbb{R}[\Theta_{\mathcal{T}}]$. This object is geometrically very simple but not invariant across $[\mathcal{T}, \Theta_{\mathcal{T}}]$.

- Finite, discrete BN models form a *subclass* of the class of staged probability trees.
- Staged trees are graphically more complex but much more expressive than BNs.
- These trees have the same algebraic properties as *Chain Event Graphs*, introduced in Smith & Anderson (2008).

POLYNOMIALS IN TREE MODELS

We define an *interpolating polynomial* of the model $\mathbb{P}_{(\mathcal{T}, \Theta_{\mathcal{T}})}$ as $c_{g, \mathcal{T}}(\theta) = \sum_{\lambda \in \Lambda(\mathcal{T})} g(\lambda) \pi_{\theta}(\lambda)$, where $g = 1$ or an indicator function. Denote by $[\mathcal{T}, \Theta_{\mathcal{T}}]^c \subseteq [\mathcal{T}, \Theta_{\mathcal{T}}]$ the class of *polynomially equivalent* tree representations, sharing the same interpolating polynomial.

There is a bijective map $\mathfrak{c} : s(c(\theta)) \mapsto (\mathcal{T}, \Theta_{\mathcal{T}})$, identifying a tree representation from a *tree-compatible* polynomial c . Interpolating polynomials are tree-compatible. In practice, \mathfrak{c} maps a certain order of summation to a corresponding graph, using $\theta(e) \mapsto e \in E$ (see below).

MAIN RESULTS

PROPOSITION For a finite and discrete BN model with decomposable DAG D and class of tree representations $[\mathcal{T}, \Theta_{\mathcal{T}}]_D$ we find that $[\mathcal{T}, \Theta_{\mathcal{T}}]^c = [\mathcal{T}, \Theta_{\mathcal{T}}]_D$ for any *clique-induced* interpolating polynomial c . As a consequence, one interpolating polynomial captures all tree representations of a decomposable BN and thus, in particular, all factorisations according to D .

THEOREM Let $(\mathcal{S}, \Theta_{\mathcal{S}}), (\mathcal{T}, \Theta_{\mathcal{T}}) \in [\mathcal{T}, \Theta_{\mathcal{T}}]$ be statistically equivalent staged trees with interpolating polynomials $c_{\mathcal{S}} \in \mathbb{R}[\Theta_{\mathcal{S}}]$ and $c_{\mathcal{T}} \in \mathbb{R}[\Theta_{\mathcal{T}}]$. Then there is a bijective map $\Phi : c_{\mathcal{S}} \mapsto c_{\mathcal{T}}$. In particular, we can transform $(\mathcal{S}, \Theta_{\mathcal{S}})$ into $(\mathcal{T}, \Theta_{\mathcal{T}})$ using a finite number of tree-compatible reorderings of the interpolating polynomial in combinations with substitutions of terms of edge probabilities.

EXAMPLE: TREE REPRESENTATIONS OF A DECOMPOSABLE BN

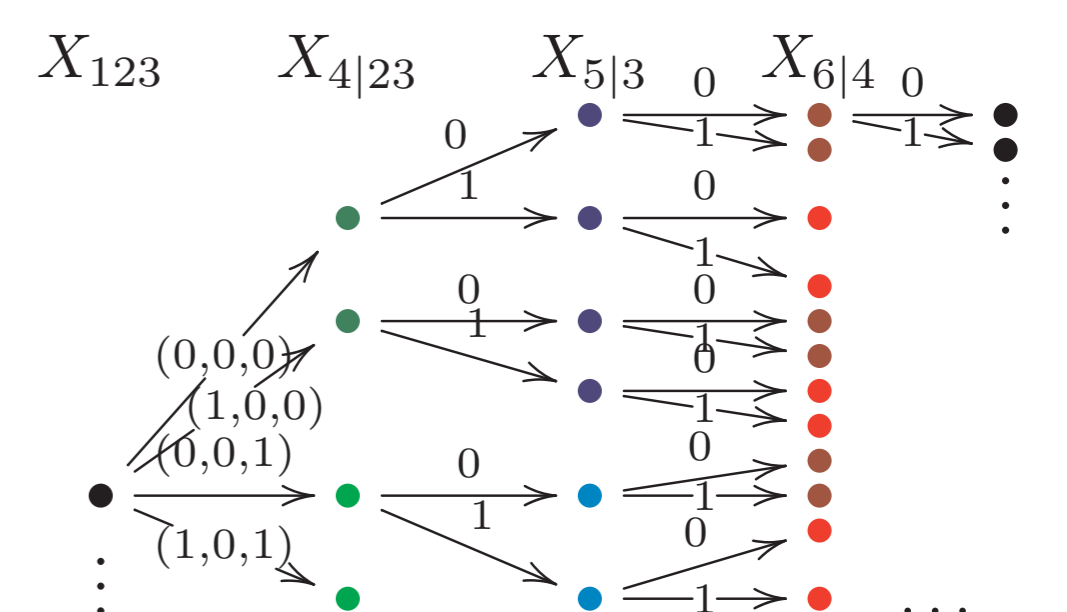
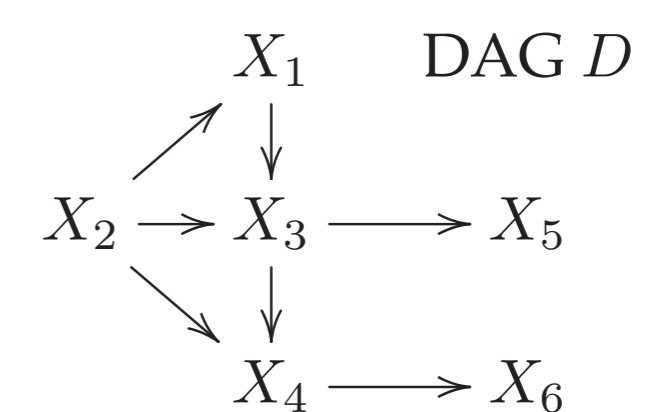
Consider a BN with (clique) parametrisation

$$p_{\theta}(x) = \theta(x_1, x_2, x_3) \theta(x_2, x_3, x_4) \theta(x_3, x_5) \theta(x_4, x_6)$$

according to a decomposable DAG D , shown on the right. Define a probability tree $(\mathcal{T}, \Theta_{\mathcal{T}}) \in [\mathcal{T}, \Theta_{\mathcal{T}}]_D$ with $\pi_{\theta}(\lambda(x)) = p_{\theta}(x)$ for all $x \in \mathbb{X}$.

The interpolating polynomial $c_{\mathcal{T}}(\theta) = \sum_{x \in \mathbb{X}} p_{\theta}(x)$ is given below. Each of the summations we see corresponds (via the map \mathfrak{c} above) to a different tree representation in the class $[\mathcal{T}, \Theta_{\mathcal{T}}]^c = [\mathcal{T}, \Theta_{\mathcal{T}}]_D$. The interpolating polynomial allows us to traverse the whole class of tree representations of the BN, and *stratified trees* are the ones giving a recursive factorisation of p_{θ} according to D .

$$\begin{aligned} c_{\mathcal{T}}(\theta) &= \sum_{(x_1, x_2, x_3) \in \mathbb{X}_{\{1,2,3\}}} \theta(x_1, x_2, x_3) \sum_{x_4 \in \mathbb{X}_4} \theta(x_2, x_3, x_4) \sum_{x_5 \in \mathbb{X}_5} \theta(x_3, x_5) \sum_{x_6 \in \mathbb{X}_6} \theta(x_4, x_6) && p_{123}(x_1, x_2, x_3) p_4(x_4 | x_2, x_3) p_5(x_5 | x_3) p_6(x_6 | x_4) \\ &= \sum_{(x_2, x_3, x_4) \in \mathbb{X}_{\{2,3,4\}}} \theta(x_2, x_3, x_4) \sum_{x_1 \in \mathbb{X}_1} \theta(x_1, x_2, x_3) \sum_{x_5 \in \mathbb{X}_5} \theta(x_3, x_5) \sum_{x_6 \in \mathbb{X}_6} \theta(x_4, x_6) && p_{234}(x_2, x_3, x_4) p_1(x_1 | x_2, x_3) p_5(x_5 | x_3) p_6(x_6 | x_4) \\ &= \sum_{(x_3, x_5) \in \mathbb{X}_{\{3,5\}}} \theta(x_3, x_5) \sum_{(x_1, x_2) \in \mathbb{X}_{\{1,2\}}} \theta(x_1, x_2, x_3) \sum_{x_4 \in \mathbb{X}_4} \theta(x_2, x_3, x_4) \sum_{x_6 \in \mathbb{X}_6} \theta(x_4, x_6) && p_{35}(x_3, x_5) p_{12}(x_1, x_2 | x_3) p_4(x_4 | x_2, x_3) p_6(x_6 | x_4) \\ &= \sum_{(x_4, x_6) \in \mathbb{X}_{\{4,6\}}} \theta(x_4, x_6) \sum_{(x_2, x_3) \in \mathbb{X}_{\{2,3\}}} \theta(x_2, x_3, x_4) \sum_{x_1 \in \mathbb{X}_1} \theta(x_1, x_2, x_3) \sum_{x_5 \in \mathbb{X}_5} \theta(x_3, x_5) && p_{46}(x_4, x_6) p_{23}(x_2, x_3 | x_4) p_{12}(x_1 | x_2, x_3) p_5(x_5 | x_3) \end{aligned}$$



CONCLUSIONS

Our paper analyses discrete and context-specific BNs as well as stratified CEGs as subclasses of staged tree models, and characterises their equivalence classes via interpolating polynomials. We developed an algorithm for the map \mathfrak{c} , identifying corresponding graphs from a tree-compatible polynomial, and have examined computational aspects of our work in Leonelli et al. (2015). Future research will focus on causal interpretations of these results, based on Cowell & Smith (2015).

REFERENCES Christiane G6rgen and Jim Q. Smith, *Equivalence Classes of Chain Event Graph Models*, in preparation.

Christiane G6rgen, Manuele Leonelli and Jim Q. Smith, *A Differential Approach for Staged Trees*, Conference Proceedings of ECSQARU 2015.

Jim Q. Smith and Paul E. Anderson, *Conditional independence and Chain Event Graphs*, Journal of Artificial Intelligence 172, 42-68, 2008.

Robert G. Cowell and Jim Q. Smith, *Causal discovery through MAP selection of stratified Chain Event Graphs*, Electronic Journal of Statistics 8, 965-997, 2014.

Peter A. Thwaites and Jim Q. Smith, *A Separation Theorem for Chain Event Graphs*, submitted to the Electronic Journal of Statistics, 2015.