



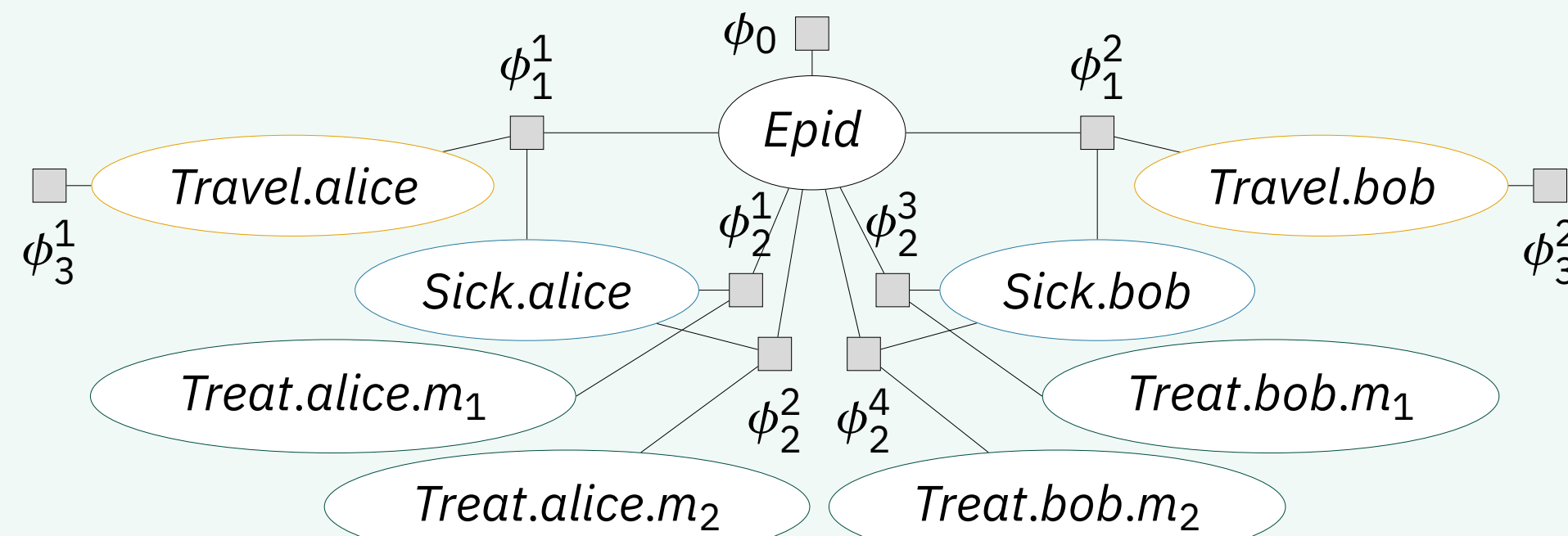
Lifted Model Construction without Normalisation: A Vectorised Approach to Exploit Symmetries in Factor Graphs

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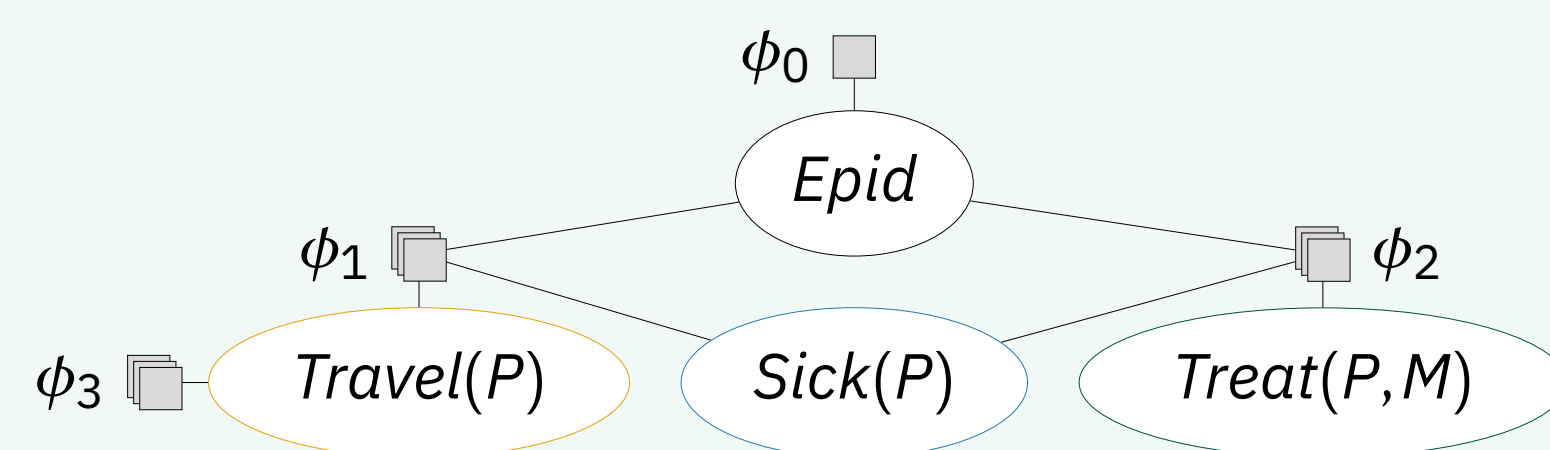
1. Motivation and Problem Setup

- Factor graphs compactly encode a distribution
- Semantics of factor graph G over a set of factors Φ :

$$P_G = \frac{1}{Z} \prod_{\phi \in \Phi} \phi$$



- Parametric factor graphs introduce logical variables to represent groups of random variables



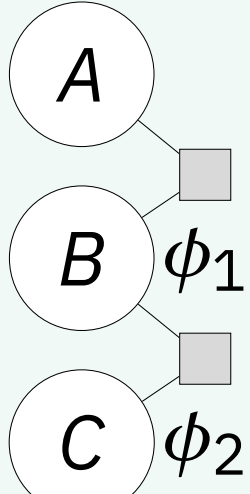
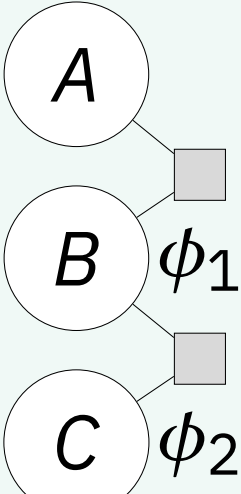
Problem Setup

Input: A factor graph G
 Output: A parametric factor graph equivalent to G
 Constraint: Handle different scaling of factors

- Factors with different scaling can occur when learning a factor graph from observed data
 - Occurrences of range values are counted
 - Normalisation is not always possible / desirable (e.g., to avoid floating point arithmetics)

2. Exchangeable Factors

- Exchangeable factors are identical up to scalar $\alpha \in \mathbb{R}^+$
- Exchangeable factors can be grouped

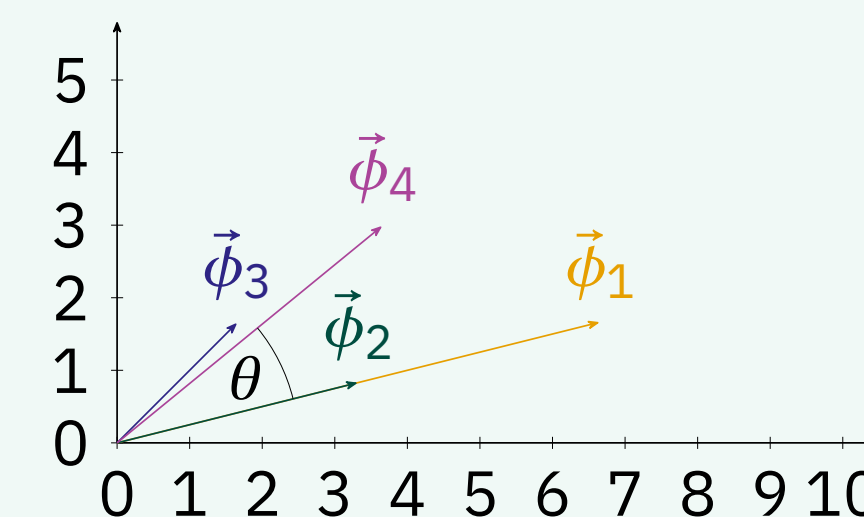
	A	B	$\phi_1(A, B)$		A	B	$\phi_1(A, B)$
	true	true	φ_1		true	true	φ_1
	true	false	φ_2		true	false	φ_2
	false	true	φ_3		false	true	φ_3
	false	false	φ_4		false	false	φ_4
	C	B	$\phi_2(C, B)$		C	B	$\phi_2(C, B)$
	true	true	φ_1		true	true	$\alpha \cdot \varphi_1$
	true	false	φ_2		true	false	$\alpha \cdot \varphi_2$
	false	true	φ_3		false	true	$\alpha \cdot \varphi_3$
	false	false	φ_4		false	false	$\alpha \cdot \varphi_4$

- Theorem:** Scaling any factor by $\alpha \in \mathbb{R}^+$ leaves the semantics of the factor graph unchanged

3. Vector Representation of Factors

- Idea: Represent table of potentials as a vector
- Theorem:** Exchangeable factors have collinear vectors

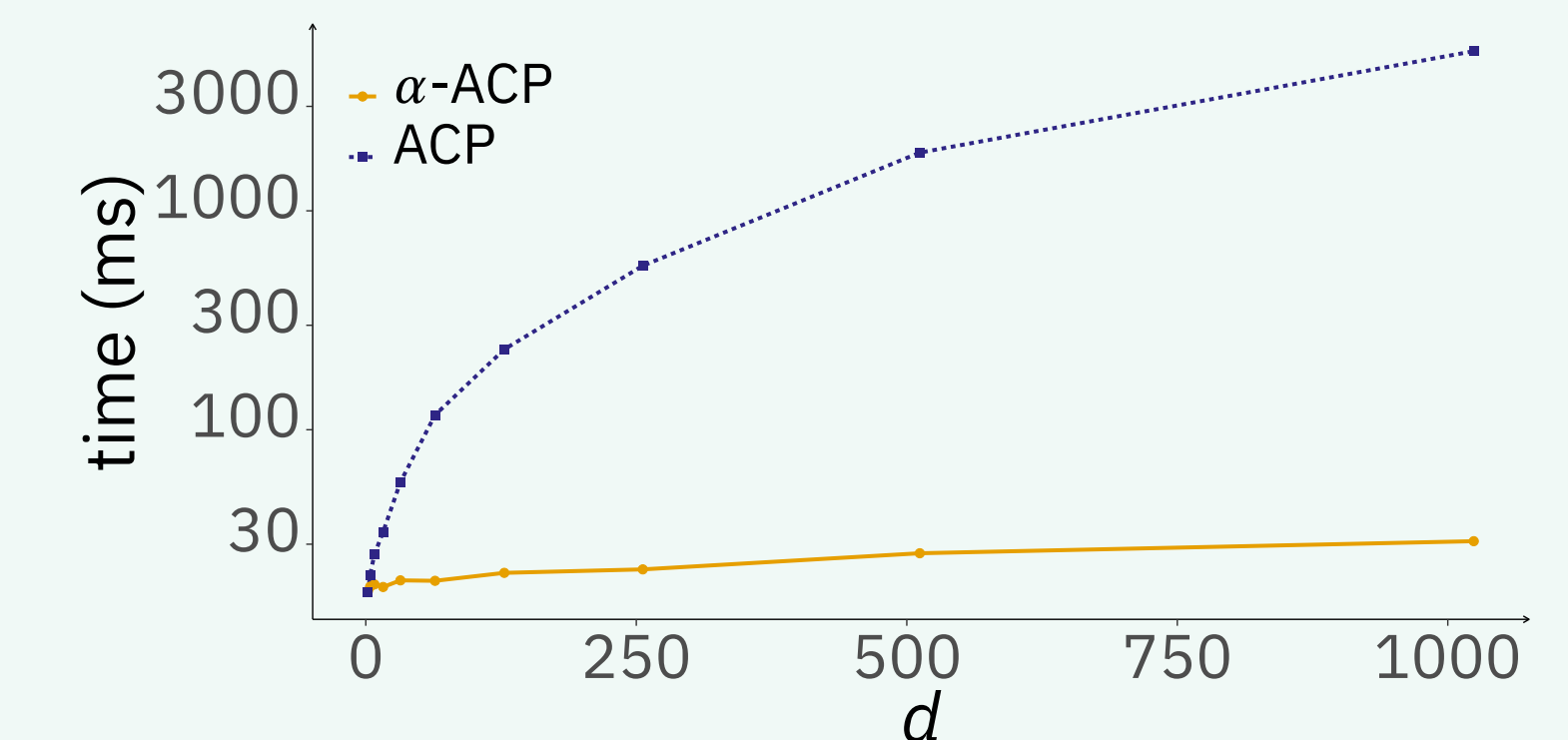
- E.g., two dimensions:
- ϕ_1, ϕ_2 are exchangeable



- Collinearity can be checked without any division operators to avoid floating point arithmetics
- Approach can also be extended:
 - To allow for permuted arguments
 - To allow for approximate exchangeability

4. Experiments

- Comparison of run times for probabilistic inference
 - ACP is the previous state-of-the-art algorithm
 - α -ACP is our generalisation of ACP
 - d controls the size of the input factor graph
- Average time over factor graphs where a proportion of $p \in \{0.01, 0.05, 0.1, 0.15\}$ of factors is scaled
- ACP is not able to detect exchangeable factors with different scales, whereas α -ACP is



- β is average number of queries after which offline overhead of α -ACP compared to ACP amortises
- Negative values for β are omitted as there is no overhead in this case (hence the missing boxes)
- After at most 10 queries, the overhead amortises

