

Lost Relatives of the Gumbel Trick

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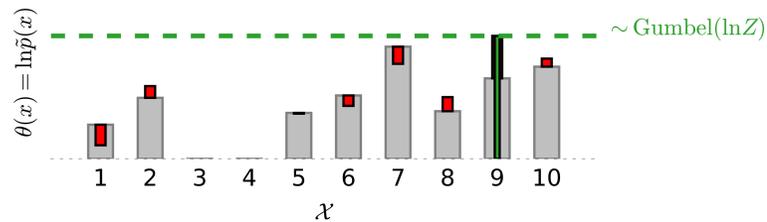
Gumbel trick

Given:

$$\tilde{\mathbf{p}} = (\tilde{p}_1, \tilde{p}_2, \dots, \tilde{p}_N) = (e^{\theta_1}, e^{\theta_2}, \dots, e^{\theta_N})$$

Trick to sample from \mathbf{p} and estimate $Z = \sum_i \tilde{p}_i$:

Perturb log-probs θ_i with Gumbel distribution and find MAP:



What's new?

New, related tricks that yield

- lower MSE estimators of Z
- new bounds on Z in discrete graphical models with negligible additional computational cost.

Competing exponential clocks



$\sim \text{Exp}(\lambda_1)$

(1) $\Pr(\text{clock } i \text{ rings first}) \propto \lambda_i$

(2) Time of first ring:

$$\min_i T_i \sim \text{Exp}\left(\sum_i \lambda_i\right)$$



$\sim \text{Exp}(\lambda_2)$



$\sim \text{Exp}(\lambda_N)$

Taking $\lambda_i = \tilde{p}_i$:

(1) $\Pr(\text{clock } i \text{ rings first}) = p_i$

(2) Time of first ring:

$$\min_i T_i \sim \text{Exp}(Z)$$

Gumbel trick: apply $g(x) = -\ln x - c$ to the clocks



...



$\sim \theta_1 + \text{Gumbel}(0)$

$\sim \theta_2 + \text{Gumbel}(0)$

$\sim \theta_N + \text{Gumbel}(0)$

(1) $\Pr(\text{maximum at } i) = p_i$

(2) Distribution of maximum value:

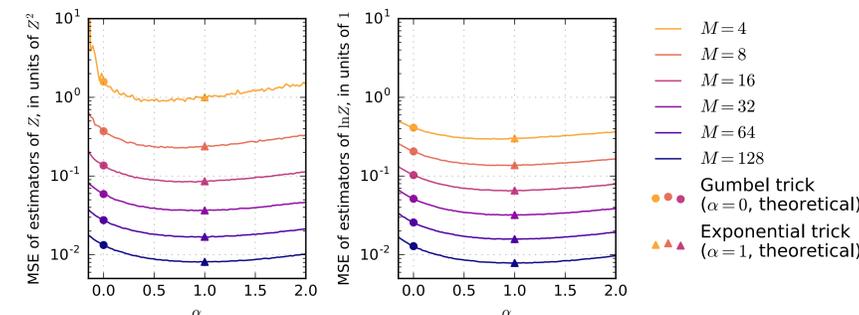
$$-\ln\left(\min_i T_i\right) - c \sim \ln Z + \text{Gumbel}(0)$$

Relatives of the Gumbel trick

Obtain them by applying different functions g to the clocks.

Trick	$g(x)$	Mean $f(Z)$	Asymptotic var. of \hat{Z}
Gumbel	$-\ln x - c$	$\ln Z$	$\frac{\pi^2}{6} Z^2$
Exponential	x	$\frac{1}{Z}$	Z^2
Weibull α	$x^\alpha, \alpha > 0$	$Z^{-\alpha} \Gamma(1 + \alpha)$	$\frac{1}{\alpha^2} \left(\frac{\Gamma(1+2\alpha)}{\Gamma(1+\alpha)^2} - 1 \right) Z^2$
Fréchet α	$x^\alpha, \alpha \in (-1, 0)$	$Z^{-\alpha} \Gamma(1 + \alpha)$	$\frac{1}{\alpha^2} \left(\frac{\Gamma(1+2\alpha)}{\Gamma(1+\alpha)^2} - 1 \right) Z^2$
Pareto	e^x	$\frac{Z}{Z-1}$ for $Z > 1$	$\frac{Z^2}{(Z-2)^2}$
Tail t	$\mathbb{1}_{\{x>t\}}$	e^{-tZ}	$\frac{(1-e^{-tZ})^2}{t^2}$

Full-rank perturbations

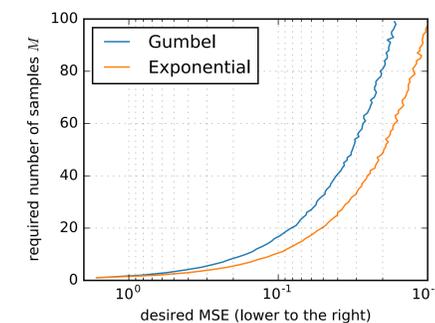


Takeaway:

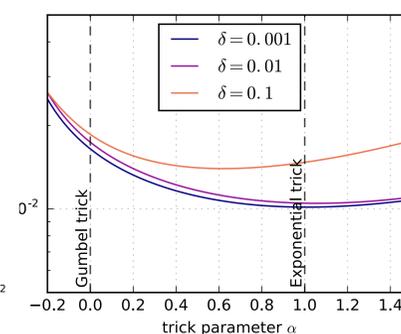
Save up to 40% samples by using the Exponential trick instead of the Gumbel trick. Applications:

- A* sampling of Maddison et al. [2014]
- Large-scale sampling of Chen and Ghahramani [2016]

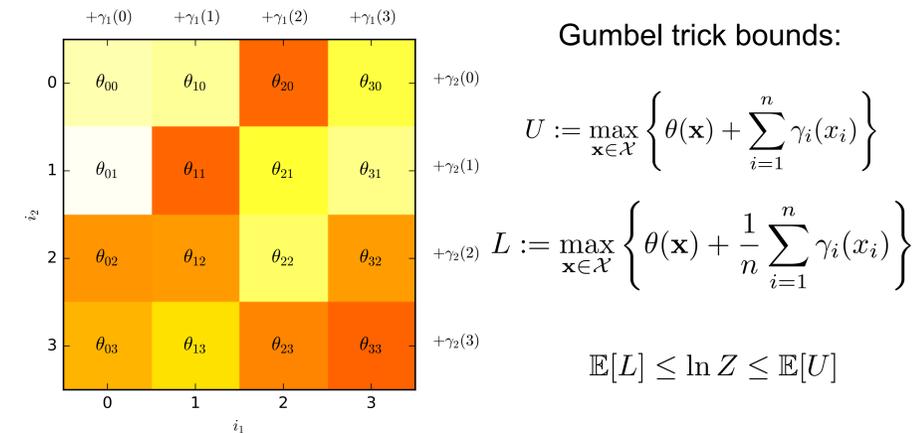
A* sampling:



MAB approximate MAP:



Discrete graphical models



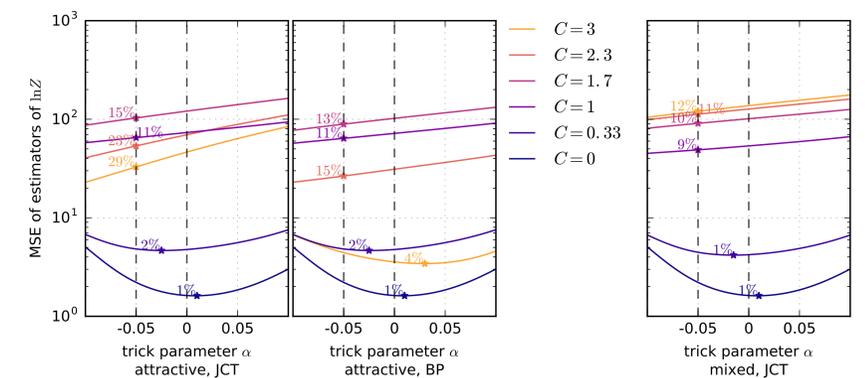
New bounds from Weibull and Fréchet tricks

For any $\alpha \in (-1, 0) \cup (0, \infty)$:

$$\ln Z \leq n \frac{\ln \Gamma(1 + \alpha)}{\alpha} + nc - \frac{1}{\alpha} \ln \mathbb{E}_\gamma [e^{-\alpha U}]$$

$$\ln Z \geq c + \frac{\ln \Gamma(1 + \alpha)}{\alpha} - \frac{1}{n\alpha} \ln \mathbb{E}[\exp(-n\alpha L)]$$

Low-rank perturbations (discrete graphical models):



Pointers

Contact: matej.balog@gmail.com

Slides: <http://matejbalog.eu/en/research/>

Code: <https://github.com/matejbalog/gumbel-relatives>