Let the Flows Tell: Solving Graph Combinatorial Optimization Problems with GFlowNets

NeurIPS 2023 spotlight

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Abstract the graph generation example...



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Emmanuel Bengio, blog of "Flow Network based Generative Models for Non-Iterative Diverse Candidate Generation"

Maximum independent set (MIS)

• Given an undirected graph g, an independent set is a set of vertices x where every pair of vertices are not neighbours



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• MIS: find largest independent set *x*

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Optimization as probabilistic inference

- $R(x | g) \leftarrow$ exponential of "number of vertices in x"
- Instead of maximize R(x | g)
- train a GFlowNet to sample prop to (tempered) reward

GFlowNet conditional on graph structure g















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Deleu, et al. Bayesian Structure Learning with Generative Flow Networks. UAI 2022.

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$$\mathscr{L}(\theta) = \frac{1}{B} \sum_{b=1}^{B} \mathscr{C}_{\text{DB}}(s^{b}, s'^{b}; \theta)$$



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- Intermediate signals
 - forward-looking (FL)



Deleu, et al. Bayesian Structure Learning with Generative Flow Networks. UAI 2022. Pan, et al. Better Training of GFlowNets with Local Credit and Incomplete Trajectories. ICML 2023.

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- Intermediate signals
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MIS results

Method	Түре	SMALL			LARGE			SATLIB		
		Size↑	$Drop \downarrow$	Тіме↓	Size↑	Drop \downarrow	Тіме↓	SIZE ↑	Drop \downarrow	Time \downarrow
Gurobi KaMIS	OR OR	$\begin{array}{c} 19.98\\ 20.10\end{array}$	$0.01\%\ 0.00\%$	47:34 1:24:12	$40.90 \\ 43.15$	$5.21\%\ 0.00\%$	2:10:26 2:03:36	$\begin{array}{c} 425.95\\ 425.96\end{array}$	$0.00\%\ 0.00\%$	3:43:19 4:15:41
PPO Intel DGL Ours	UL SL SL UL	19.01 18.47 17.36 19.18	$5.42\%\ 8.11\%\ 13.61\%\ 4.57\%$	1:17 13:04 12:47 0:32	32.32 34.47 34.50 37.48	25.10% 20.12% 20.05% 13.14%	7:33 20:17 23:54 04:22	421.49 423.54	1.05% 0.57%	13:12 23:13

Other CombOpt problems

Other problem results

METHOD	Τγρε	МС			MDS			MCUT		
		Size↑	$DROP \downarrow$	TIME↓	Size↓	$Gap \downarrow$	Тіме↓	Size↑	$DROP \downarrow$	Time↓
Gurobi SDP	OR OR	19.05 —	0.00%	1:55	27.89 —	0.00%	1:47	732.47 700.36	$0.00\%\ 4.38\%$	13:04 35:47
GREEDY MFA Erdos Anneal Ours	H H UL UL UL	13.53 14.82 12.02 14.10 16.24	$\begin{array}{c} 28.98\% \\ 22.15\% \\ 36.90\% \\ 25.98\% \\ 14.75\% \end{array}$	0:25 0:27 0:41 0:41 0:42	37.39 36.36 30.68 29.24 28.61	$\begin{array}{c} 25.41\%\\ 23.29\%\\ 9.09\%\\ 4.62\%\\ 2.52\%\end{array}$	2:13 2:56 1:00 1:01 2:20	688.31 704.03 693.45 696.73 704.30	6.03% 3.88% 5.33% 4.88% 3.85%	0:13 1:36 0:46 0:45 2:57

Thank you for listening!

https://github.com/zdhNarsil/Awesome-GFlowNets