Supplementary Information: A Framework of Hierarchical Attacks to Network Controllability

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Emails: Y. Lou (felix.lou@my.cityu.edu.hk); L. Wang (wanglin@sjtu.edu.cn); G. Chen (eegchen@cityu.edu.hk) Source code of this work is available in: https://fylou.github.io/sourcecode.html

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1 Controllability Curves Under Various Attack Simulations

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Figure S28: Edge-removal attacks on N = 500 ER



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2 Generation Methods for Complex Networks

Nine typical directed synthetic network models are adopted for simulation, namely the Erdös–Rényi random graph (ER) [1], Newman–Watts small-world (SW) network [2], generic scale-free (SF) network [3–5], *q*-snapback network (QS) [6], *q*-snapback with redirected edges (QR) [7], random triangle (RT) network [8], and random rectangle (RR) network [8], extremely homogeneous (HO) network [9], and onion-like (OL) network [10]. The HO networks are empirically with the optimal controllability robustness [9].

The detailed generation methods and parameter settings of the nine synthetic networks are introduced as following.

2.1 ER: Erdös–Rényi Random Graph Networks

An ER network is generated as follows:

- 1. Start with N isolated nodes.
- 2. Pick up all possible pairs of nodes from the N given nodes, denoted by i and j ($i \neq j, i, j = 1, 2, ..., N$), once and once only. Connect each pair of nodes by a directed edge with probability $p_{RG} \in [0, 1]$, where the edge has the same probability directing from i to j, or j to i.

Given the numbers of N and M, let $p_{RG} = \frac{M}{N(N-1)}$. To exactly control the number of generated edges to be M, uniformly-randomly adding or removing edges can be performed. Here, when adding an edge, the direction can be random.

2.2 SW: Newman–Watts Small-world Networks

An SW network is generated as follows:

- 1. Start with a directed N-node loop having K connected nearest-neighbors on each side of each node.
- 2. Additional edges with random directions are added without removing any existing edges.

Set K = 2 in the following, namely, a node *i* is connected to its two nearest neighbors on each side, with nodes i - 1, i + 1, i - 2 and i + 2, via edges $A_{i-1,i}$, $A_{i,i+1}$, $A_{i-2,i}$ and $A_{i,i+2}$.



Figure S55: Node-removal attacks on real-world networks

2.3 SF: Scale-Free Networks

An SF network is generated as follows:

- 1. Start with N isolated nodes.
- 2. A weight $w_i = (i + \theta)^{-\sigma}$ is assigned to node *i*, with $\sigma \in [0, 1)$ and $\theta \ll N$.
- 3. Two nodes *i* and *j* ($i \neq j, i, j = 1, 2, ..., N$) are randomly picked from the pool with a probability proportional to the weights w_i and w_j , respectively. Then, an edge A_{ij} from *i* to *j* is added (if the two nodes are already connected, do nothing).
- 4. Repeat Step 3), until M edges have been added.

The resulting network has a power-law distribution $k^{-\gamma}$ with $\gamma = 1 + \frac{1}{\sigma}$, where k is the degree variable, which is independent of θ . Here, σ is set to 0.999, and thus $\gamma = 2.001$.

2.4 QS: q-Snapback Networks

Consider a q-snapback network (QS) with only one layer r_{QS} for simplicity. This QS is generated as follows:

1. Start with a directed chain of N nodes, where each node i (i = 1, 2, ..., N - 1) has an edge $A_{i,i+1}$.



Figure S56: Edge-removal attacks on real-world networks

2. For each node $i = r_{QS} + 1$, $r_{QS} + 2$,..., N, it connects backward to the previously-appeared nodes $i - l \times r_{QS}$ ($l = 1, 2, ..., \lfloor i/r_{QS} \rfloor$), with the same probability $q \in [0, 1]$.

In the following experimental study, r_{QS} is set to 2. Given N = 1000 and M = 5000, q is estimated to be 0.008 for fair comparisons. To exactly generate M edges, uniformly-randomly edge-adding with random direction should be applied.

2.5 QR: q-Snapback Networks with Redirected Edges

Consider a QR with only one layer r_{QR} for simplicity. This QR is generated as follows:

- 1. Start with a directed chain of N nodes, where each node i (i = 1, 2, ..., N 1) has an edge $A_{i,i+1}$.
- 2. For each node $i = r_{QR} + 1$, $r_{QR} + 2$,..., N, it connects backward to the previously-appeared nodes $i l \times r_{QR}$ ($l = 1, 2, ..., \lfloor i/r_{QR} \rfloor$), with the same probability $q \in [0, 1]$. With a probability p_{re} , this snapback edge is redirected.

In the following experimental study, r_{QS} is set to 2. Given N = 1000 and M = 5000, q is estimated to be 0.008 for fair comparisons. To exactly generate M edges, uniformly-randomly edge-adding with random direction should be applied. In the experiments, p_{re} is set to 0.5.

2.6 RT: Random Triangle Networks

Triangular structure, which has been observed benefit to the robustness of controllability [6] and network stability [11,12], is frequently observed in real-life situations.

A directed random triangle network (RTN) is generated as follows:

- 1. Start with N 3 isolated nodes, with the other 3 nodes connected in a directed triangle.
- 2. Randomly pick up two nodes, *i* and *j*, without edge A_{ij} or A_{ji} (otherwise, do nothing). Then, randomly pick up a node *k* from all the neighbors of node *j*. If there is an edge A_{jk} , then add two edges A_{ij} and A_{ki} ; otherwise (e.g., with an edge A_{kj}), add two edges A_{ji} and A_{ik} .
- 3. Repeat Step 2), until M edges have been added.

2.7 RR: Random Rectangle Networks

The above directed RTN is extended to a random rectangle network (RRT), as follows:

- 1. Start with N 4 isolated nodes, and the other 4 nodes are connected in a directed rectangle.
- 2. Randomly pick up three nodes, *i*, *j* and *k*, without edges between any pair of them (otherwise, do nothing). Then, randomly pick up a node *w* from the neighbors of node *k*. If there is an edge A_{kw} , then add edges A_{wi} , A_{ij} , and A_{jk} ; otherwise (e.g., with an edge A_{wk}), add edges A_{ki} , A_{ij} , and A_{jw} .
- 3. Repeat Step 2), until M edges have been added.

2.8 HO: Extremely Homogeneous Networks

The in- and out-degree distributions of a directed HO network satisfy the following condition:

$$\lfloor M/N \rfloor \le k_i^{in,out} \le \lceil M/N \rceil, \ i = 1, 2, \dots, N,$$
(1)

where N is the number nodes; M is the number edges; $k_i^{in,out}$ means both in- and out-degrees, in which as a standard notation the floor function $\lfloor x \rfloor$ returns the greatest integer less than or equal to x, and the ceiling function $\lceil x \rceil$ returns the least integer greater than or equal to x.

An HO network is generated as follows:

- 1. Given an ER network.
- 2. Perform random edge rectification (RER) until both the in- and out-degree distributions satisfy Eq. (1).

The random edge rectification (RER) operator is performed as follows: For any node i, if its in- or out-degree does not satisfy Eq. (1), edge rectification is needed. There are four possible edge rectification operations:

- 1. If $k_i^{out} < \lfloor M/N \rfloor$, then find another node k with out-degree greater than $\lceil M/N \rceil$, and randomly pick one of its out-edges, $A_{k,l}$. Delete this edge $A_{k,l}$ and add an edge $A_{i,l}$. This increases k_i^{out} by one and decreases k_k^{out} by one.
- 2. If $k_i^{out} > \lceil M/N \rceil$, then randomly pick one of its out-edges $A_{i,j}$, and find another node k with out-degree less than $\lfloor M/N \rfloor$. Delete this edge $A_{i,j}$ and add an edge $A_{k,j}$. This decreases k_i^{out} by one and increases k_k^{out} by one.
- 3. If $k_i^{in} < \lfloor M/N \rfloor$, then find another node k with in-degree greater than $\lceil M/N \rceil$, and randomly pick one of its in-edges $A_{l,k}$. Delete this edge $A_{l,k}$ and add an edge $A_{l,i}$. This increases k_i^{in} by one and decreases k_k^{in} by one.
- 4. If $k_i^{in} > \lceil M/N \rceil$, then randomly pick one of its in-edges $A_{j,i}$, and find another node k with in-degree less than $\lfloor M/N \rfloor$. Delete this edge $A_{j,i}$ and add an edge $A_{j,k}$. This decreases k_i^{in} by one and increases k_k^{in} by one.

2.9 OL: Onion-like Networks

An OL network is generated as follow:

- 1. Given an SF network.
- 2. Perform random edge-swapping with degree reservation [10]. If the *connected robustness* measure improves after swapping, then keep it; otherwise, discard the swapping. Until the *connected robustness* measure stagnates.

The degree distribution of the resultant OL follows the same power-law distribution as the SF network.

3 Comparison of Overall Controllability

Table 1: Comparison of attack strategies on the nine synthetic networks (N = 500), where B represents betweenness; D represents degree; C represents closeness; R represents random; Hy represents hybrid; IC represents initial critical edges; HB represents hierarchical betweenness; HD represents hierarchical degree; HC represents hierarchical closeness; HR represents hierarchical random.

N=500				Node	e Attack			Edge Attack					
		HB/B	HD/D	HC/C	HR/R	HB/Hy	HD/Hy	HB/B	HD/D	HR/R	HB/Hy	HD/Hy	HR/IC
$\langle k \rangle$ =3	ER	1.184	1.138	1.241	1.490	1.090	1.092	1.120	1.584	1.377	1.125	0.892	1.255
	SW	1.251	1.093	1.196	1.291	1.035	1.060	1.157	1.350	1.435	1.264	0.977	1.440
	SF	1.036	1.011	1.030	1.309	1.005	1.007	1.165	1.257	1.177	1.191	1.111	1.075
	QS	1.268	1.200	1.504	1.408	1.103	1.181	1.120	2.252	1.462	1.118	0.878	1.300
	QR	1.289	1.172	1.291	1.470	1.108	1.135	1.114	1.699	1.443	1.112	0.867	1.398
	RT	1.155	1.105	1.134	1.632	1.056	1.061	1.254	1.704	1.401	1.286	1.078	1.228
	RR	1.223	1.120	1.228	1.564	1.081	1.089	1.216	1.688	1.441	1.222	1.005	1.297
	HO	1.240	1.192	1.410	1.427	1.141	1.145	1.146	1.351	1.375	1.092	1.134	1.374
	OL	1.039	1.016	1.027	1.306	1.008	1.011	1.162	1.264	1.171	1.189	1.114	1.066
	ER	1.196	1.209	1.315	1.518	1.135	1.120	1.064	1.432	1.324	1.066	0.651	1.265
	SW	1.257	1.179	1.311	1.429	1.104	1.126	1.068	1.389	1.294	1.138	0.737	1.320
	SF	1.052	1.026	1.051	1.429	1.011	1.013	1.200	1.372	1.221	1.221	1.098	1.089
	QS	1.226	1.253	1.657	1.390	1.175	1.182	1.072	2.067	1.423	1.064	0.714	1.399
$\langle k \rangle = 5$	QR	1.265	1.185	1.338	1.507	1.136	1.140	1.060	1.413	1.320	1.067	0.659	1.339
	RT	1.176	1.169	1.191	1.623	1.096	1.102	1.129	1.519	1.332	1.154	0.750	1.318
	RR	1.239	1.165	1.301	1.576	1.119	1.124	1.110	1.435	1.327	1.108	0.739	1.315
	HO	1.213	1.188	1.366	1.427	1.174	1.149	1.070	1.408	1.239	1.037	0.986	1.241
	OL	1.052	1.025	1.048	1.420	1.013	1.014	1.211	1.385	1.226	1.228	1.101	1.092
<pre></pre>	ER	1.204	1.203	1.373	1.441	1.180	1.128	1.058	1.416	1.221	1.047	0.473	1.233
	SW	1.209	1.208	1.309	1.414	1.146	1.113	1.025	1.350	1.143	1.123	0.501	1.084
	SF	1.074	1.057	1.092	1.722	1.025	1.039	1.213	1.784	1.387	1.234	0.884	1.221
	QS	1.169	1.311	1.982	1.286	1.174	1.151	1.014	1.805	1.356	1.260	0.589	1.389
	QR	1.240	1.199	1.320	1.376	1.161	1.109	1.030	1.338	1.158	1.100	0.500	1.192
	RT	1.183	1.193	1.302	1.486	1.125	1.112	1.033	1.321	1.306	1.047	0.514	1.290
	RR	1.192	1.188	1.293	1.450	1.118	1.136	1.062	1.305	1.137	1.107	0.538	1.191
	HO	1.212	1.208	1.375	1.337	1.190	1.148	1.067	1.367	1.122	1.192	0.753	1.140
	OL	1.075	1.038	1.085	1.711	1.023	1.029	1.209	1.688	1.382	1.211	0.870	1.182

N=1500		Node Attack							Edge Attack					
		HB/B	HD/D	HC/C	HR/R	HB/Hy	HD/Hy	HB/B	HD/D	HR/R	HB/Hy	HD/Hy	HR/IC	
(k)=3	ER	1.286	1.177	1.322	1.534	1.134	1.144	1.139	1.587	1.448	1.135	0.863	1.352	
	SW	1.290	1.082	1.221	1.334	1.065	1.086	1.163	1.327	1.423	1.215	0.925	1.433	
	SF	1.033	1.010	1.029	1.258	1.006	1.006	1.157	1.235	1.161	1.160	1.101	1.058	
	QS	1.314	1.213	1.880	1.387	1.082	1.199	1.131	2.237	1.463	1.123	0.879	1.241	
	QR	1.323	1.171	1.361	1.484	1.130	1.158	1.125	1.694	1.440	1.128	0.873	1.347	
	RT	1.161	1.102	1.178	1.638	1.061	1.075	1.258	1.694	1.399	1.285	1.080	1.213	
	RR	1.250	1.123	1.261	1.579	1.085	1.094	1.230	1.706	1.426	1.236	1.023	1.273	
	HO	1.283	1.184	1.478	1.471	1.155	1.172	1.152	1.335	1.374	1.092	1.142	1.374	
	OL	1.034	1.011	1.030	1.260	1.005	1.006	1.150	1.232	1.154	1.150	1.098	1.055	
	ER	1.228	1.206	1.349	1.586	1.151	1.158	1.049	1.550	1.436	1.055	0.649	1.374	
	SW	1.309	1.201	1.347	1.491	1.150	1.175	1.077	1.421	1.422	1.138	0.692	1.428	
	SF	1.045	1.021	1.046	1.354	1.009	1.012	1.203	1.359	1.220	1.200	1.103	1.098	
	QS	1.260	1.283	2.433	1.394	1.176	1.230	1.071	2.269	1.517	1.073	0.705	1.505	
$\langle k \rangle = 5$	QR	1.273	1.199	1.367	1.553	1.159	1.162	1.065	1.513	1.439	1.063	0.645	1.450	
	RT	1.206	1.154	1.234	1.723	1.104	1.117	1.144	1.566	1.453	1.179	0.726	1.381	
	RR	1.243	1.144	1.298	1.621	1.117	1.132	1.118	1.450	1.442	1.125	0.711	1.432	
	HO	1.240	1.189	1.450	1.537	1.167	1.159	1.086	1.452	1.408	1.026	0.986	1.404	
	OL	1.045	1.019	1.047	1.359	1.009	1.011	1.204	1.340	1.218	1.207	1.107	1.087	
$\langle k \rangle$ =10	ER	1.234	1.219	1.380	1.547	1.166	1.129	1.040	1.446	1.456	1.037	0.416	1.466	
	SW	1.268	1.226	1.416	1.562	1.164	1.134	1.037	1.411	1.382	1.049	0.421	1.387	
	SF	1.065	1.036	1.081	1.556	1.020	1.025	1.214	1.617	1.352	1.219	0.945	1.180	
	QS	1.202	1.339	4.723	1.344	1.229	1.218	1.027	2.060	1.552	1.104	0.484	1.557	
	QR	1.254	1.221	1.418	1.600	1.177	1.150	1.038	1.438	1.405	1.036	0.422	1.397	
	RT	1.239	1.196	1.312	1.624	1.152	1.140	1.058	1.414	1.497	1.066	0.461	1.501	
	RR	1.241	1.182	1.349	1.636	1.134	1.139	1.064	1.441	1.380	1.071	0.492	1.373	
	HO	1.218	1.192	1.418	1.539	1.159	1.121	1.046	1.384	1.384	1.080	0.671	1.400	
	OL	1.069	1.037	1.083	1.565	1.022	1.026	1.210	1.610	1.366	1.210	0.934	1.179	

Table 2: Comparison of attack strategies on the nine synthetic networks (N = 500), where B represents betweenness; D represents degree; C represents closeness; R represents random; Hy represents hybrid; IC represents initial critical edges; HB represents hierarchical betweenness; HD represents hierarchical degree; HC represents hierarchical closeness; HR represents hierarchical random.

4 Other Supplementary Materials



(a) Results of node attacks on ER, HO, and OL (N = 1000): hierarchical(b) Results of node attacks on ER, HO, and OL (N = 1000): hierarchical betweenness-based (N-HB) and betweenness-based (N-B). degree-based (N-HD) and degree-based (N-D).



(c) Results of node attacks on ER, HO, and OL (N = 1000): random(d) Results of node attacks on ER, HO, and OL (N = 1000): three hierar-(N-R), hierarchical random (N-HR) and initial critical (N-IC). chical attacks (N-HB, N-HD and N-HR) and hybrid (N-Hy).

Figure S57: Node-removal attacks on ER, HO, and OL (N = 1000).

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Figure S58: Number of critical nodes (boxplots) and initial controllability (stars *) against the average degree of (a) SF and (b) OL networks.

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