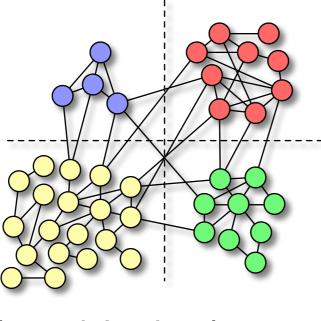
GrasP: Distributed Streaming Graph Partitioning

Casey Battaglino, Robert Pienta, Richard Vuduc Georgia Institute of Technology

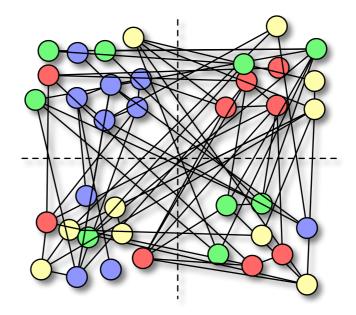
KDD HPGM '15

Contributions

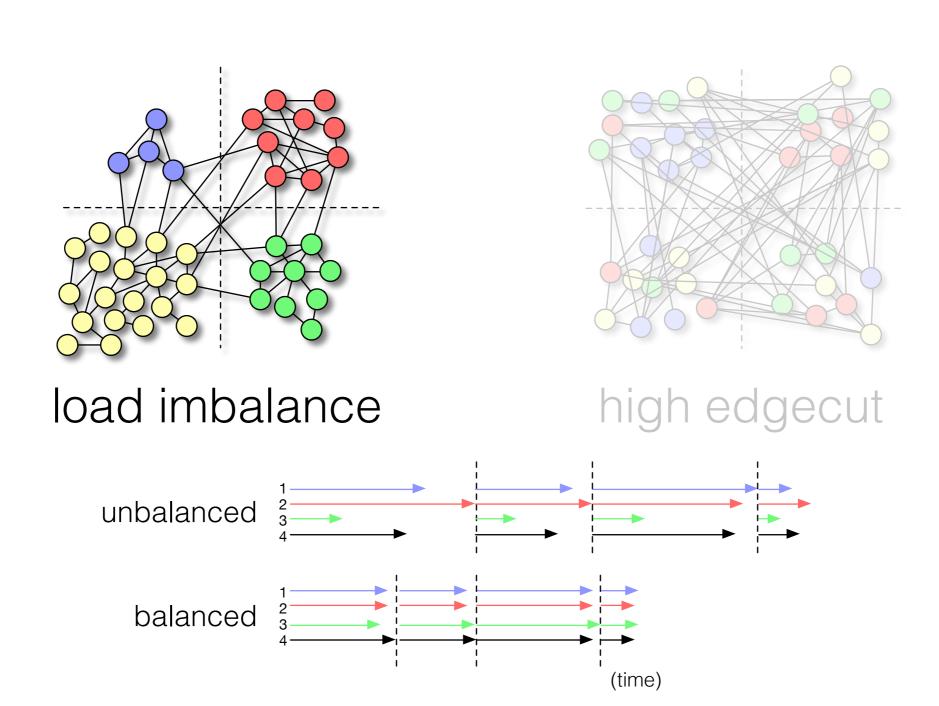
- Parallel implementation of re-streaming partitioning algorithm
- GraSP interface is easily substituted for existing HPC partitioners such as ParMETIS
- Performance evaluation on large graphs on a state-of-the-art machine, with very favorable comparison to existing methods

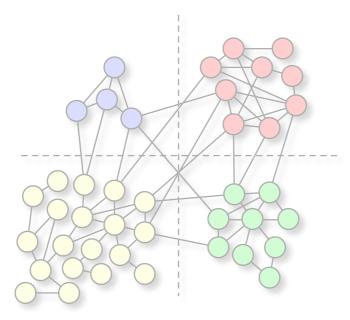


load imbalance



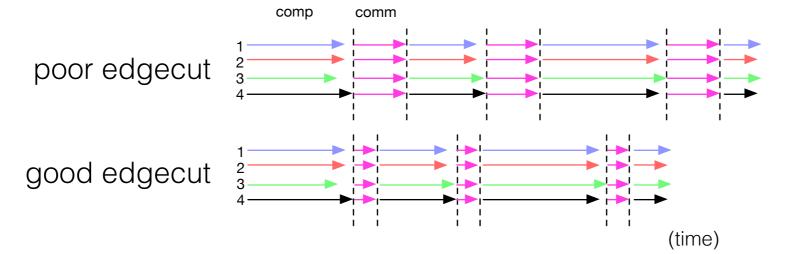
high edgecut

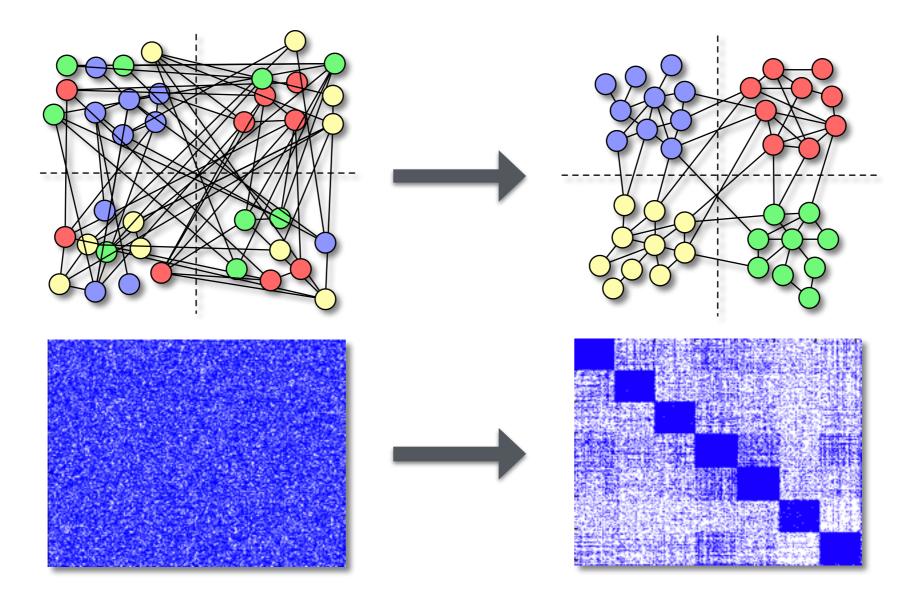




load imbalance

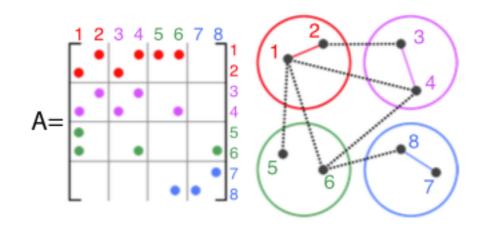
high edgecut





(spy plot)

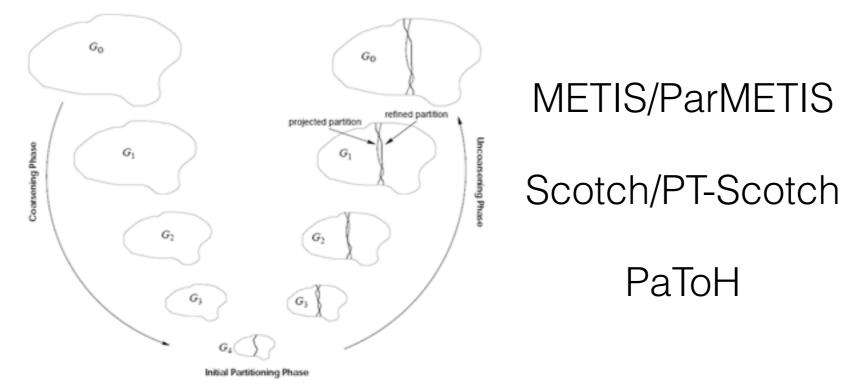
Graph Partitioning



BFS/Shortest Path Connectivity PageRank Betweenness Centrality etc. etc.

HPC: Solvers, SpMV

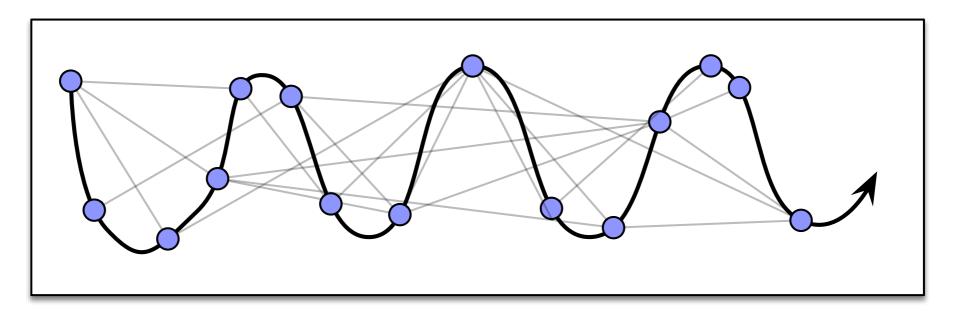
HPC Graph Partitioners



- Offline, generally divide-and-conquer ('multilevel')
- Achieve excellent partitions for wide variety of graphs
- Suitable for moderate-sized graphs, but heavy parallel overhead
- Re-partitioning is slow

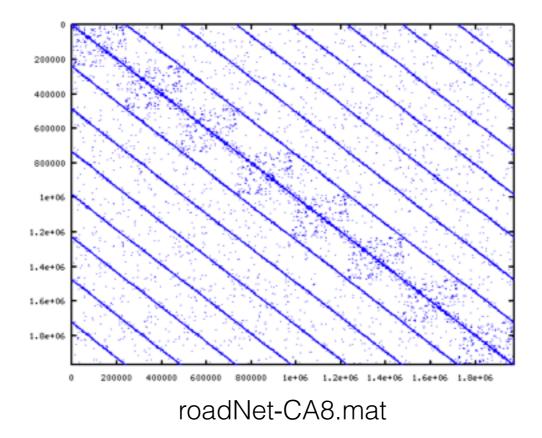
algorithmic alternatives: spectral, geometric, graph growing, random walk

Streaming Graph Partitioners



- Stream over vertices in any order
- Touch each vertex once
- Simple to implement and very fast
- Quality partitions for low-diameter/scale-free graphs
- Suitable for re-partitioning / dynamic partitioning

Streaming Graph Partitioners



Note: Struggle on higher diameter graphs

Streaming Partitioning [SK2012, TKRV2012]

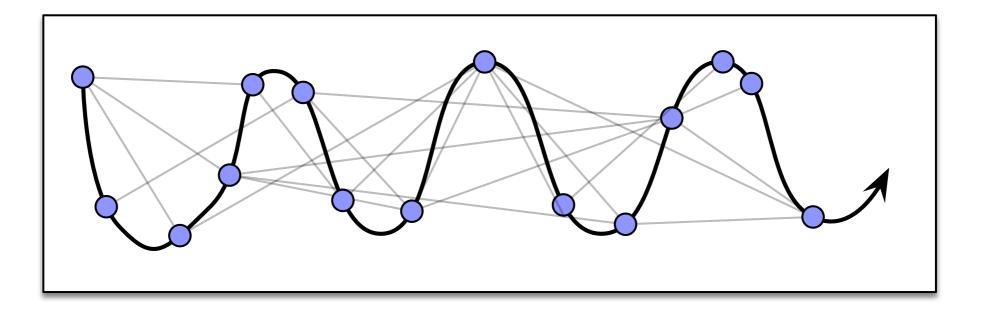
Set all P_i to \emptyset ; **foreach** $v \in V(G)$ as it arrives at time t **do** $\begin{vmatrix} j \leftarrow \arg\max_{i \in \{1,...,p\}} |P_i^t \cap N(v)| - \alpha \frac{\gamma}{2} |P_i^t|^{\gamma-1}; \\ \text{Add } v \text{ to set } P_j^{t+1}; \end{vmatrix}$

$$|P_1| = 4$$

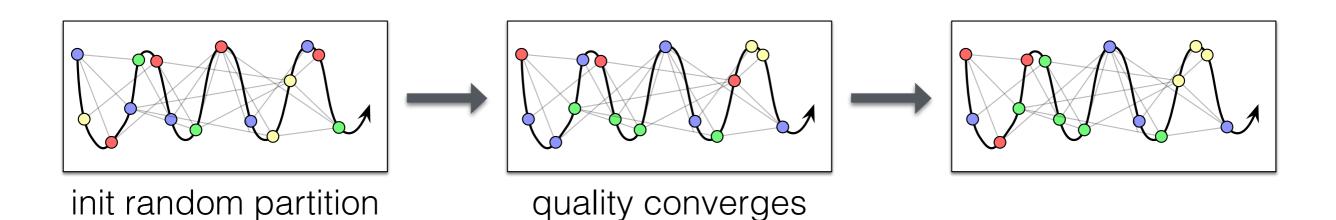
end

 $|P_1 \cap N(v)| = 2$

Algorithm 1: Serial streaming FENNEL partitioner



Restreaming Partitioning [NU2013]



temper parameters over time to favor balance

GraSP: Parallel Restreaming

```
for each process p do in parallel

vorder \leftarrow rand\_perm(\{0, ..., |V(G_{local})|\});
```

Randomly assign local vertices to partitions $P_{i,p}^0$;

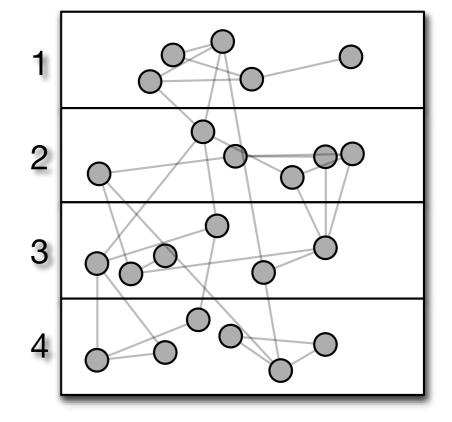
end

```
for run \leftarrow \{1 \dots n_s\} do

for each process p do in parallel

\begin{vmatrix} for each process p do in parallel \\ for each <math>v \in vorder do

\begin{vmatrix} j \leftarrow \arg\max_{i \in \{1,\dots,p\}} |P_i^t \cap N(v)| - \alpha \frac{\gamma}{2} |P_i^t|^{\gamma-1}; \\ Add v \text{ to set } P_{j,p}^{t+1}; \\ end \\ end \\ MPL_{ALLGATHER} global partition assignments; \\ \alpha \leftarrow t_{\alpha} \alpha \\ end \\ end \\ \end{vmatrix}
```



Initialization

```
for each process p do in parallel

vorder \leftarrow rand\_perm(\{0, ..., |V(G_{local})|\});

Randomly assign local vertices to partitions P_{i,p}^{0};

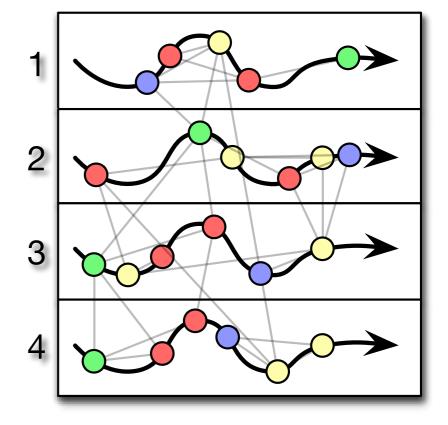
end
```

```
for run \leftarrow \{1 \dots n_s\} do

for each process p do in parallel

for each v \in vorder do

\begin{vmatrix} j \leftarrow \arg\max |P_i^t \cap N(v)| - \alpha \frac{\gamma}{2} |P_i^t|^{\gamma-1}; \\ Add v \text{ to set } P_{j,p}^{t+1}; \\ end \\ MPLALLGATHER global partition assignments; \\ \alpha \leftarrow t_{\alpha} \alpha \\ end \\ end \\ \end{vmatrix}
```



Multiple Runs

for each process p do in parallel $vorder \leftarrow rand_perm(\{0, ..., |V(G_{local})|\});$ Randomly assign local vertices to partitions $P_{i,p}^{0};$ end

for $run \leftarrow \{1 \dots n_s\}$ do

```
for each process p do in parallel

for each v \in vorder do

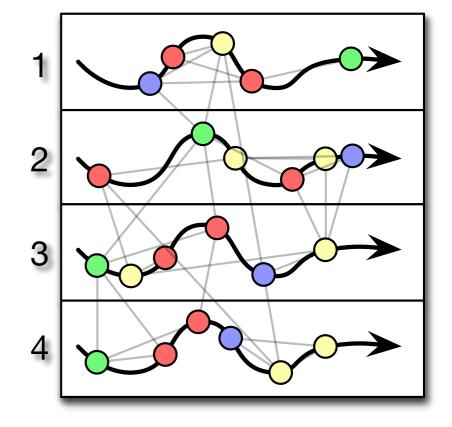
\begin{vmatrix} j \leftarrow \arg\max_{i \in \{1,...,p\}} |P_i^t \cap N(v)| - \alpha \frac{\gamma}{2} |P_i^t|^{\gamma-1}; \\ Add v \text{ to set } P_{j,p}^{t+1}; \\ end

end

MPLALLGATHER global partition assignments;

\alpha \leftarrow t_{\alpha} \alpha
```

\mathbf{end}



Stream

for each process p do in parallel $vorder \leftarrow rand_perm(\{0, ..., |V(G_{local})|\});$ Randomly assign local vertices to partitions $P_{i,p}^{0};$ end

```
for run \leftarrow \{1 \dots n_s\} do

for each process p do in parallel

foreach v \in vorder do

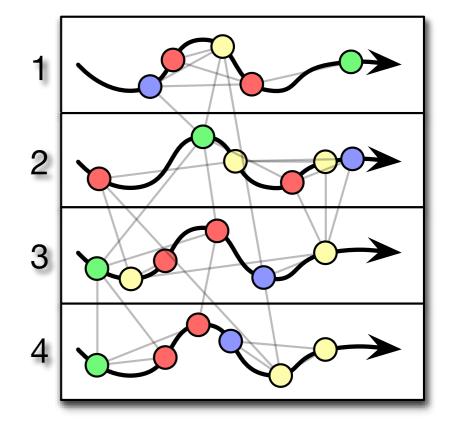
\begin{vmatrix} j \leftarrow \arg\max |P_i^t \cap N(v)| - \alpha \frac{\gamma}{2} |P_i^t|^{\gamma-1}; \\ Add v \text{ to set } P_{j,p}^{t+1}; \\ end

end

MPLALLGATHER global partition assignments;

\alpha \leftarrow t_{\alpha} \alpha
```

\mathbf{end}



Communicate

for each process p do in parallel $| vorder \leftarrow rand_perm(\{0, ..., |V(G_{local})|\});$ Randomly assign local vertices to partitions $P_{i,p}^{0};$ end

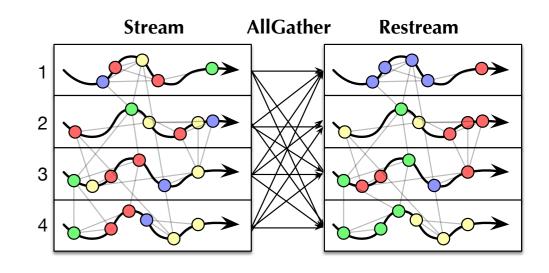
for $run \leftarrow \{1 \dots n_s\}$ do

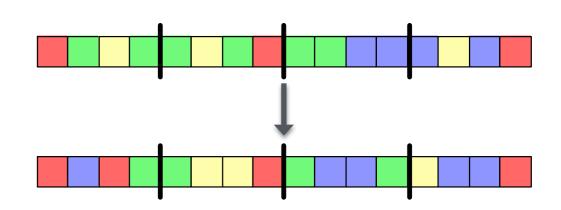
```
for each process p do in parallel

for each v \in vorder do

\begin{vmatrix} j \leftarrow \arg\max_{i \in \{1,...,p\}} |P_i^t \cap N(v)| - \alpha \frac{\gamma}{2} |P_i^t|^{\gamma-1}; \\ Add v \text{ to set } P_{j,p}^{t+1}; \\ end \\ end \\ MPLALLGATHER global partition assignments; \\ \alpha \leftarrow t_{\alpha} \alpha
```

\mathbf{end}





```
for each process p do in parallel
      vorder \leftarrow rand\_perm(\{0, \ldots, |V(G_{local})|\});
     Randomly assign local vertices to partitions P_{i,p}^0;
end
for run \leftarrow \{1 \dots n_s\} do
     for each process p do in parallel
           for
each v \in vorder do
       \begin{vmatrix} j \leftarrow \underset{i \in \{1, \dots, p\}}{\operatorname{argmax}} |P_i^t \cap N(v)| - \alpha \frac{\gamma}{2} |P_i^t|^{\gamma - 1}; \\ \text{Add } v \text{ to set } P_{j, p}^{t+1}; \end{vmatrix}
                                                                                             post: final partition computed
           end
     end
     MPI_ALLGATHER global partition assignments;
     \alpha \leftarrow t_{\alpha} \alpha
end
```

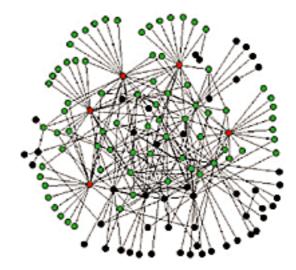
System

- Edison @ NERSC
- 5576 Compute Nodes
- Two 12-core "Ivy Bridge" processors per node
- Cray Aries interconnect
- MPI v3.0



Real-World Data: SNAP Data Sets⁺

Synthetic Data: R-MAT

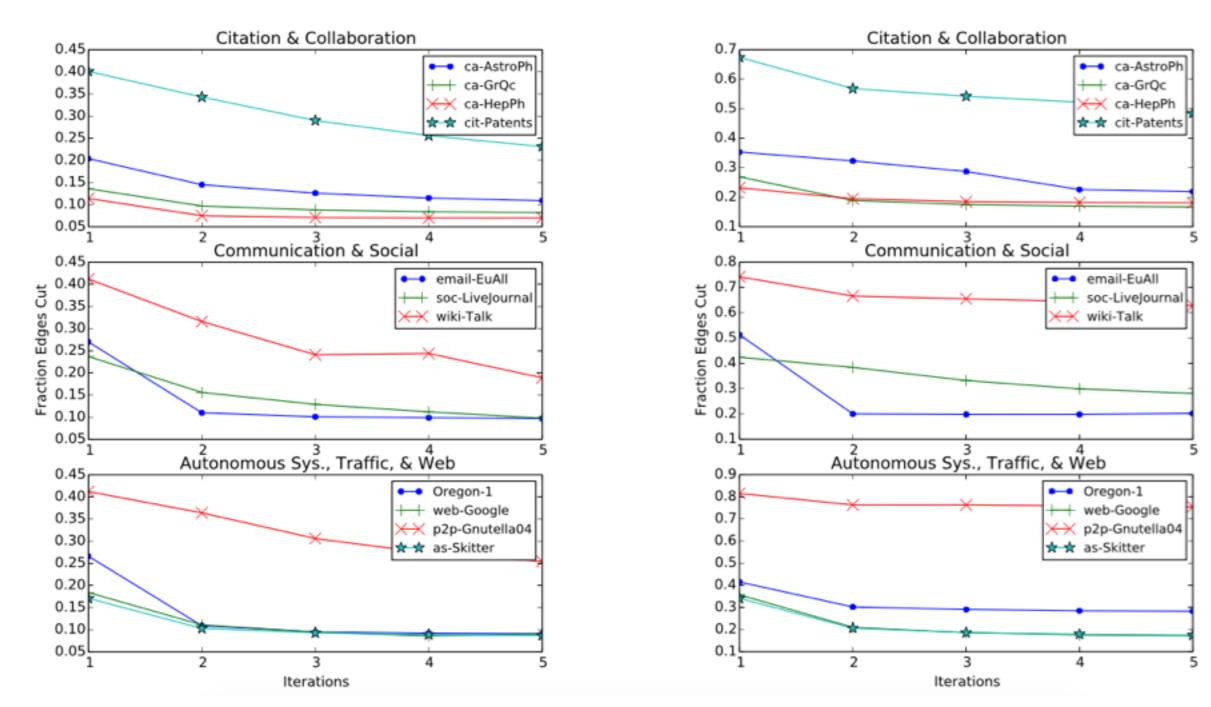


+: <u>https://snap.stanford.edu/data/</u>

Verification on real-world data sets

Data Set	N	nnz	$\lambda_{p=2}$	$\lambda_{p=8}$
soc-LiveJournal	4,847,571	$68,\!993,\!773$	0.234	0.463
as-Skitter	1,696,415	22,190,596	0.166	0.324
cit-Patents	3,774,768	16,518,948	0.402	0.726
roadNet-CA	1,971,281	5,533,214	0.186	0.360
web-Google	916,428	5,105,039	0.189	0.336
wiki-Talk	2,394,385	5,021,410	0.411	0.752
amazon0302	262,111	1,234,877	0.202	0.370
soc-Slashdot0902	82,168	948,464	0.236	0.382
ca-AstroPh	18,772	396,160	0.232	0.413
cit-HepPh	34,546	421,578	0.343	0.646
email-EuAll	265,214	420,045	0.280	0.538
Oregon-1	11,492	46,818	0.224	0.406
p2p-Gnutella04	10,879	39,994	0.415	0.747

Verification on real-world data sets



Synthetic graphs: R-MAT

Scale	26	27	28	29	30	31
V(G)	$67 \mathrm{M}$	134M	268M	537M	1.07B	2.15B
$ \mathbf{E}(\mathbf{G}) $	1.07B	2.14B	4.29B	8.58B	17.1B	34.3B

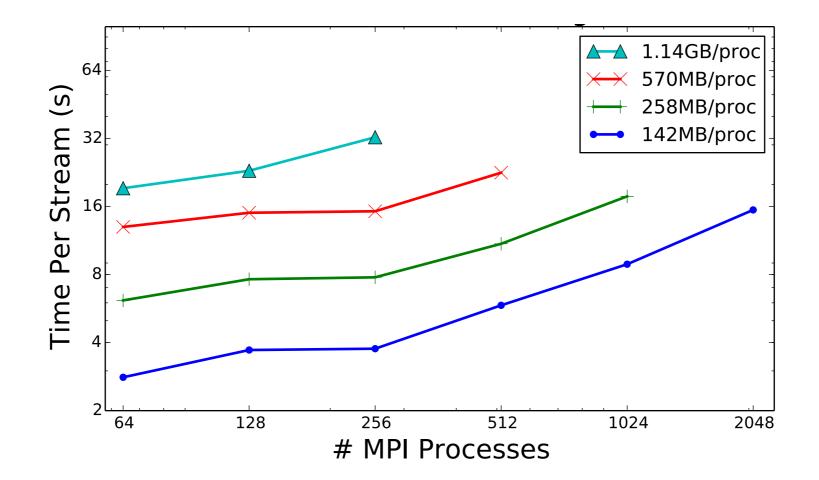
GraSP vs ParMETIS: Scale-22 R-MAT⁺

#procs	λ_{metis}	λ_{grasp}	$t_{metis}(s)$	$t_{grasp}(s)$
8	0.36	0.29	307.8	0.72
16	0.38	0.41	221.9	0.45
32	0.40	0.54	194.9	0.31

We generally encountered performance ~3 order of magnitude better.

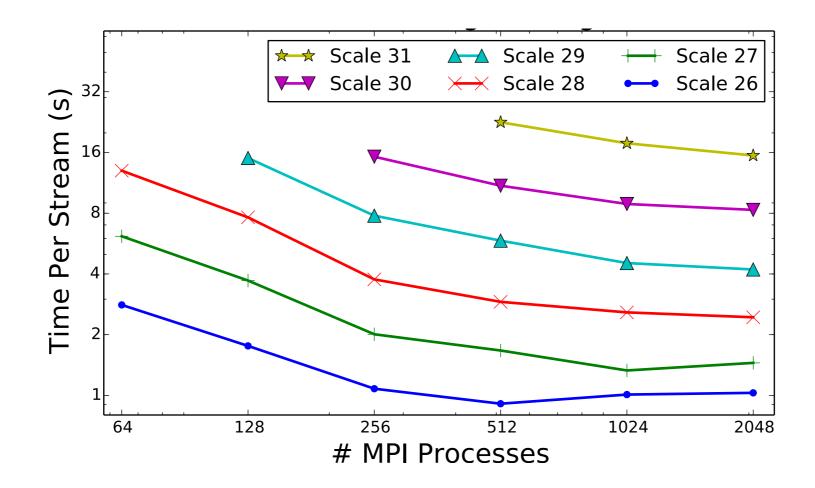
To save core-hours we didn't run ParMETIS above Scale-22

+: 4194304 nodes, 67108864 edges

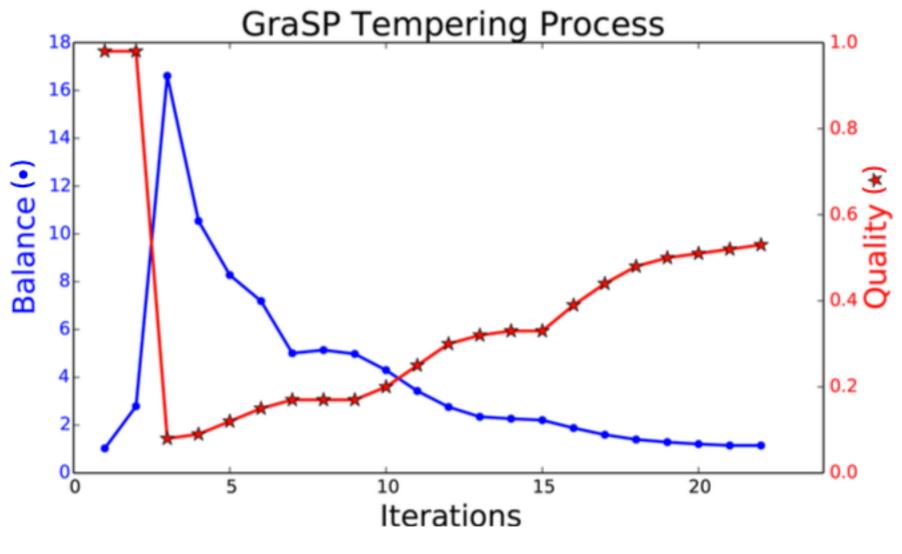


Weak Scaling

Scale	26	27	28	29	30	31
V(G)	67M	134M	268M	537M	1.07B	2.15B
E(G)	1.07B	2.14B	4.29B	8.58B	17.1B	34.3B

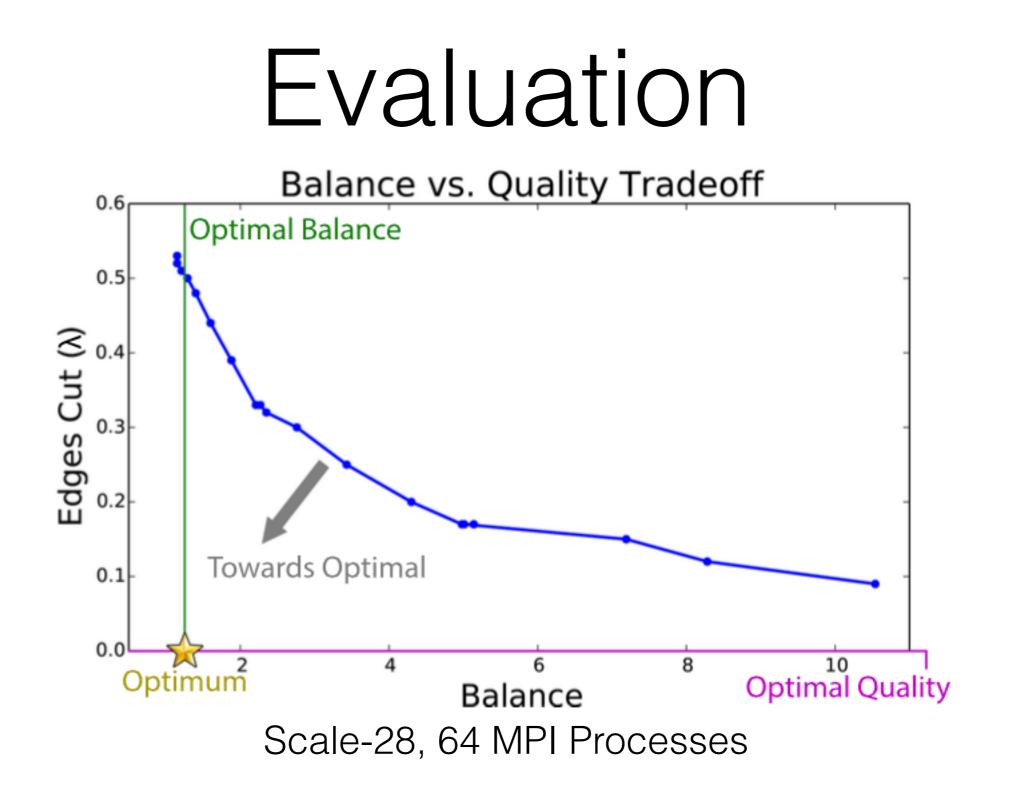


Strong Scaling



Scale-28, 64 MPI Processes

"Tempering"



"Tempering"

Conclusion

- Streaming partitioning is simple, scalable and effective (for the right kinds of graphs)
- Streaming partitioning can operate orders of magnitude faster than sophisticated parallel partitioners with similar quality
- Streaming partitioning deserves more attention from the HPC community