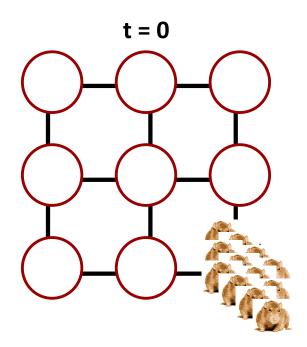
Assignment 3: GraphRats



Topics

- Application
- Implementation Issues
- Optimizing for Sequential Performance
- Optimizing for Parallel Performance
- Useful Advice

Basic Idea



Graph

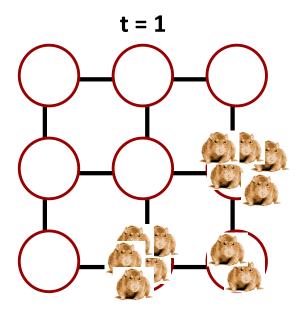
K X K grid

Initial State

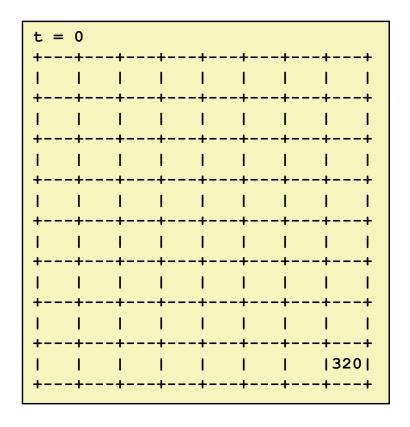
Start with all R rats in corner

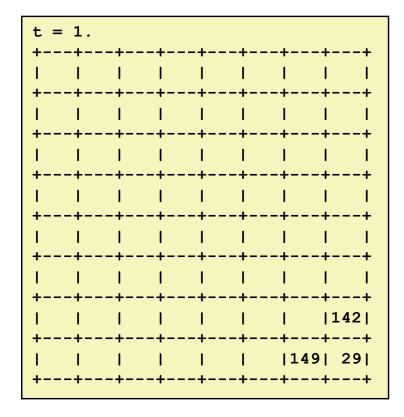
Transitions

- Each rat decides where to move next
 - Don't like crowds
 - But also don't like to be alone
- Weighted random choice

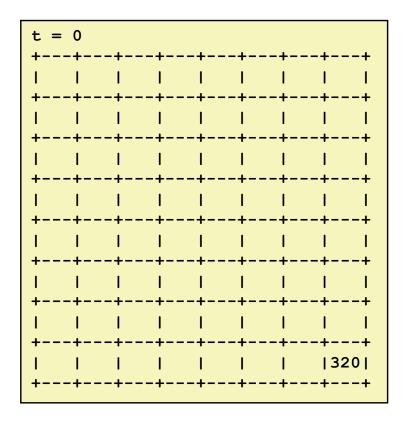


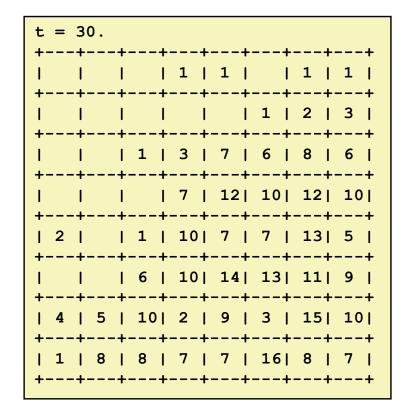
Node Count Representation





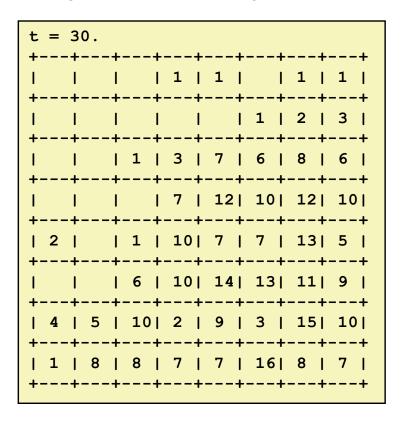
Simulation Example



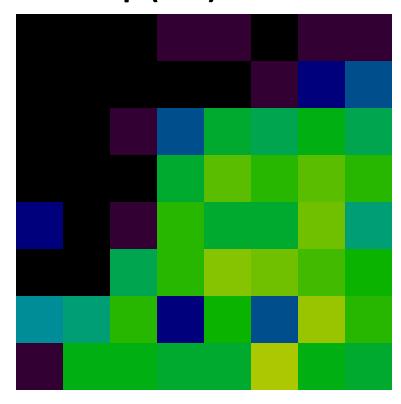


Visualizations

Text ("a" for ASCII)



Heat Map ("h")



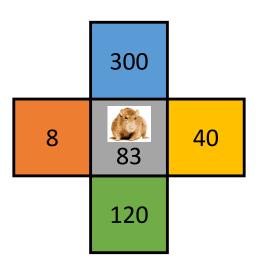
Running it yourself

```
linux> cd some directory
linux> git clone https//github.com/cmu15418/asst3-s18.git
linux> cd asst3-s18/code
Linux> make demoX
     X from 1 to 10
```

Demos

- 1: Text visualization, synchronous updates
- 2: Heap-map, synchronous updates

Determining Rat Moves



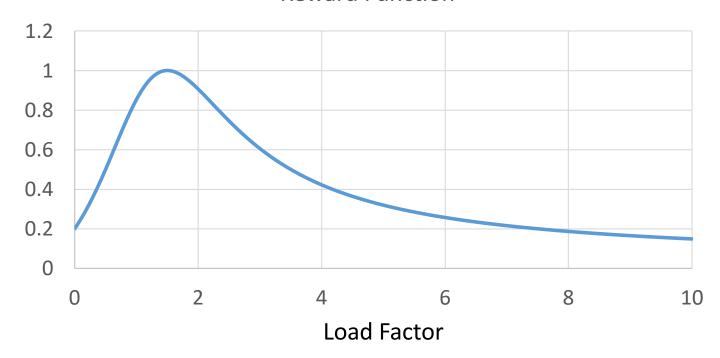
- Count number of rats at current and adjacent locations
 - Adjacency structure represented as graph
- Compute reward value for each location
 - Based on load factor l = count/average count
 - Ideal load factor (= 1.5)
 - α Fitting parameter (= 0.5)

Reward(
$$l$$
) = $\frac{1}{1 + (\log_2 [1 + \alpha(l - l^*)])^2}$

Reward Function

Reward(
$$l$$
) = $\frac{1}{1 + (\log_2 [1 + \alpha(l - l^*)])^2}$

Reward Function

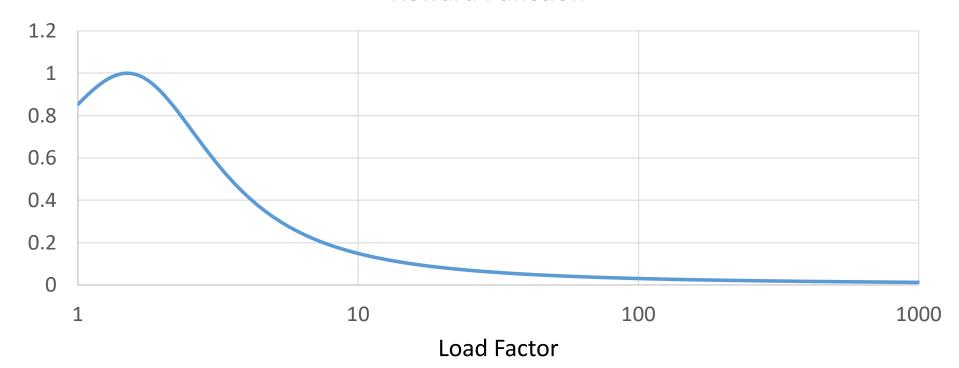


- Maximized at load factor 1.5
 - Just above average population
 - Drops for smaller loads (too few) and larger loads (too crowded)

Reward Function (cont.)

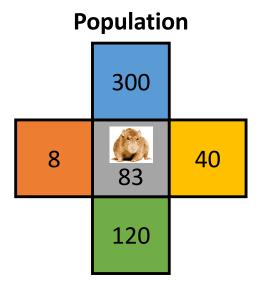
Reward(
$$l$$
) = $\frac{1}{1 + (\log_2 [1 + \alpha(l - l^*)])^2}$

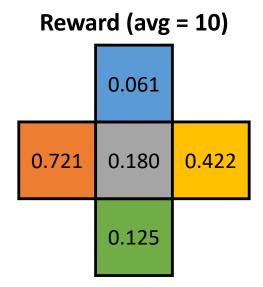
Reward Function

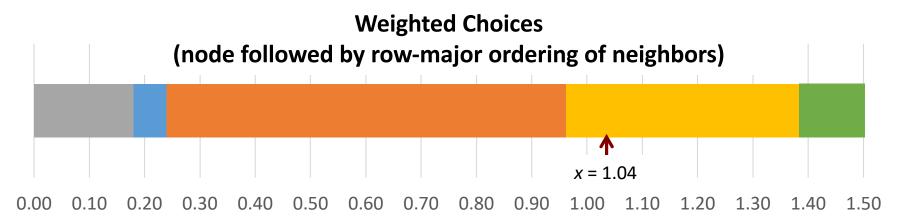


- Falls off gradually
 - Reward(1000) = 0.0123

Selecting Next Move







- Choose random number between 0 and sum of rewards
- Move according to interval hit

Update Models

Synchronous

- Demo 2
- Compute next positions for all rats, and then move them
- Causes oscillations/instabilities

Rat-order

- Demo 3
- For each rat, compute its next position and then move it
- Smooth transitions, but costly

Batch

- Demo 4
- For each batch of B rats, compute next moves and then move them
- \blacksquare B = 0.02 * R
- Smooth enough, with better performance possibilities

What We Provide

- Python version of simulator
 - Demo 4
 - Very slow
- C version of simulator
 - Faster, but still too slow
 - Demo 5: 8X8 grid, 320 rats
 - Demo 6: 160X160 grid, 1,024,000 rats
 - That's what we'll be using for benchmarks!
 - You'll have to be patient using the starter code
- Generate visualizations by piping C simulator output into Python simulator
 - Operating in visualization mode
 - See Makefile for examples

Correctness

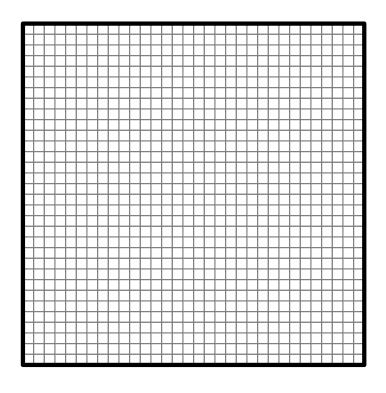
Simulator is Deterministic

- Global random seed
- Random seeds for each rat
- Process rats in fixed order

You Must Preserve Exact Same Behavior

- Python simulator generates same result as C simulator
- Use regress.py to check
- Don't rely only on the small cases it currently checks

Graphs: Grid (Demo 6)

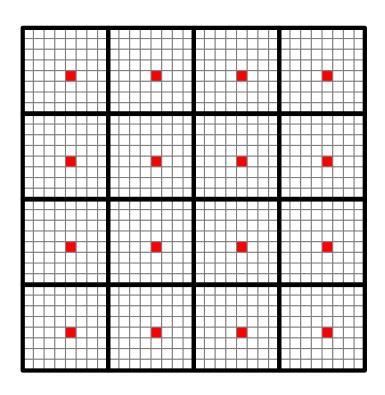


Highly regular Low degree Rats spread slowly

- k X k nodes, each with nearest neighbor connectivity
- Max degree = 4

■ k = 160: 25,600 nodes 101,760 edges

Graphs: Tiled (Demo 7)

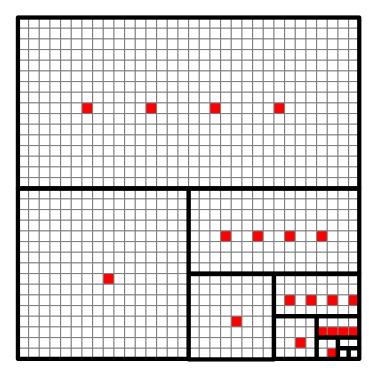


Globally regularly, locally irregular
Low/medium degree degree
Rats spread quickly within tile, slowly between tiles
Hub nodes typically have high rat counts

■ Smaller d X d regions, each with *hub* node

- Node with edge to every other node in region
- Regions connected only by nearest-neighbor edges at border
- Maximum degree = d²-1 (= 99 for benchmark graph)
- k = 160, d = 10: 25,600 nodes 152,850 edges

Fractal Graph (Demo 8)



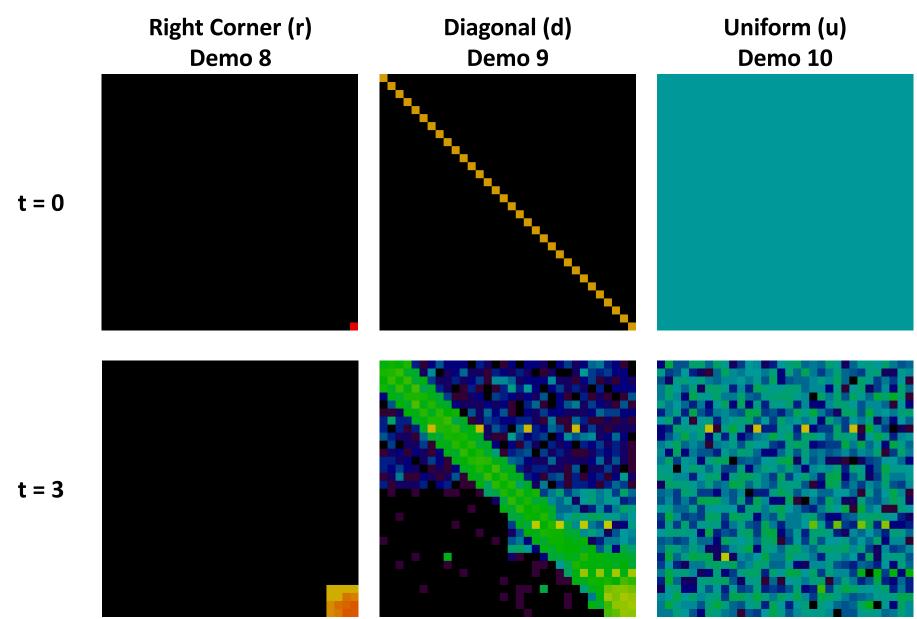
Highly irregular
Widely varying degrees
Rats spread quickly through regions
Hub nodes typically have high rat counts

Regions of varying size

- Some have 4 hub nodes, others 1
- Maximum degree = $k^2/2-1$ (= 12,799 for benchmark graph)

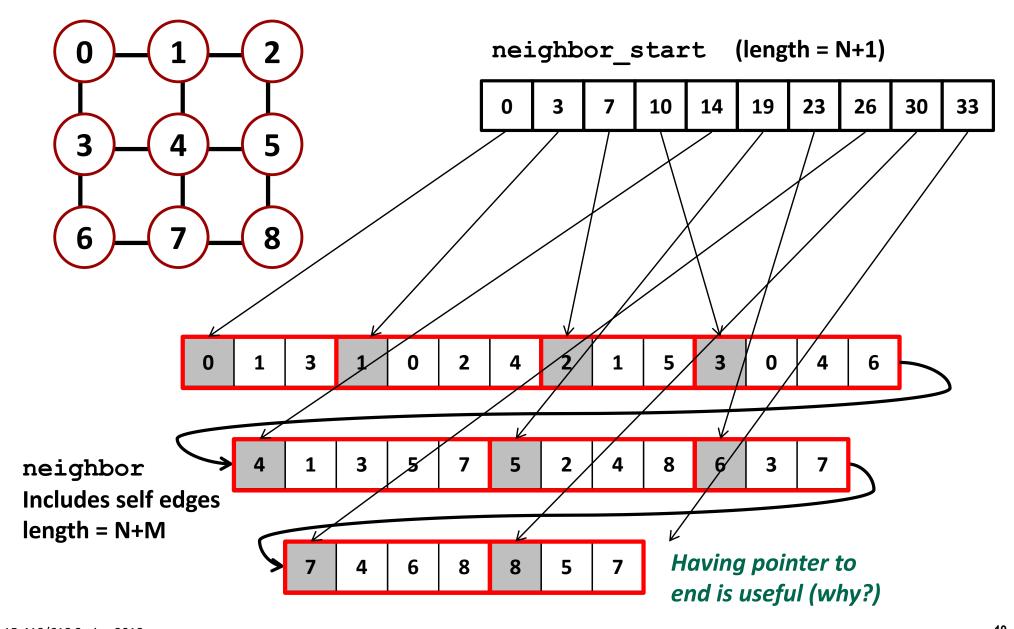
■ k = 160: 25,600 nodes 254,940 edges

States (Fractal Graph)



Graph Representation

N node, M edges



Sample Code

- From sim.c
- Compute reward value for node

```
/* Compute weight for node nid */
static inline double compute_weight(state_t *s, int nid)
{
   int count = s->rat_count[nid];
   return mweight((double) count/s->load_factor);
}
```

- Simulation state stored in state_t struct
- Reward function computed by mweight

Sample Code

- From sim.c
- Compute sum of reward values for node

```
/* Compute sum of weights in region of nid */
static inline double compute_sum_weight(state_t *s, int nid)
{
    graph_t *g = s->g;
    double sum = 0.0;
    int eid;
    int eid_start = g->neighbor_start[nid];
    int eid_end = g->neighbor_start[nid+1];
    int *neighbor = g->neighbor;
    for (eid = eid_start; eid < eid_end; eid++) {
        sum += compute_weight(s, neighbor[eid]);
    }
    return sum;
}</pre>
```

Sample Code

Compute next move for rat

```
static inline int next random move(state t *s, int r)
   int nid = s->rat position[r];
   random t *seedp = &s->rat seed[r];
   double tsum = compute sum weight(s, nid);
   graph t *g = s->g;
   double val = next_random_float(seedp, tsum);
   double psum = 0.0;
   int eid;
   int eid start = g->neighbor start[nid];
   int eid end = g->neighbor start[nid+1];
    int *neighbor = g->neighbor;
   for (eid = eid start; eid < eid end; eid++) {</pre>
       psum += compute weight(s, neighbor[eid]);
       if (val < psum) {</pre>
           return neighbor[eid];
```

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Sequential Efficiency Considerations

- Consider move computation for rat at node with degree D
 - How many (on average) iterations of loop in next random move?
 - How many calls are made to mweight?
- Suppose there are X rats in batch at single node with degree D
 - How many (on average) iterations of loop in next random move?
 - How many calls are made to mweight?

Finding Parallelism

Sequential constraints

- Must complete time steps sequentially
- Must complete each batch before starting next

Sources of parallelism

- Over nodes
 - Computing reward functions
- Over rats (within a batch)
 - Computing next moves
 - Updating node counts

Performance Targets

Mega-Rats Per Second (MRPS)

- R rats running for S steps
- Requires time T
- MRPS = $10^{-6} * R * S / T$

Runs

- 5 combinations of graph/initial state
- Compute geometric mean of MRPS's

Target performance

| | 1 Thread | 12 Threads | Speedup |
|-------------|----------|------------|---------|
| Synchronous | 32 | 256 | 8.0 |
| Batch | 20 | 70 | 3.5 |

Some Suggestions

Focus initially on sequential performance

- But think of ways that will allow parallelism
- Simple ideas / data structures generally work best
- Use timing to guide optimizations

Synchronous mode easier to make fast

Both sequentially and parallel

Machines

- Can develop on any machine
 - GHC machines work well
- Performance will be evaluated on Latedays machines
 - Batch submission process
 - These have different characteristics from GHC machines when measuring parallel performance

Instrumenting Your Code

- How do you know how much time each activity takes?
 - Create simple library using cycletimer code
 - Bracket steps in your code with library calls
 - Use macros so that you can disable code for maximum performance

```
START_ACTIVITY(ACTIVITY_NEXT);
#pragma omp parallel for schedule(static)
for (ri = 0; ri < local_count; ri++) {
    int rid = ri + local_start;
    s->rat_position[rid] = fast_next_random_move(s, rid);
}
FINISH_ACTIVITY(ACTIVITY_NEXT);
```

Evaluating Your Instrumented Code

1 thread

```
1.0 %
         194 ms
                               startup
        2077 ms
                     11.1 %
                               compute weights
        4029 ms
                     21.6 %
                               compute sums
                     62.8 %
                               find moves
       11733 ms
                     3.5 %
         651 ms
                               set ops
                     0.0 %
           3 ms
                               unknown
        25600
                           40
                                          400
                                                           1
                                                                    18.70
                                                                              21.91
+++
                                                  b
```

12 threads

```
3.2 %
 192 ms
                       startup
 426 ms
             7.0 %
                       compute weights
 940 ms
            15.5 %
                       compute sums
3168 ms
            52.3 %
                       find moves
            21.9 %
1325 ms
                       set ops
   2 ms
            0.0 %
                       unknown
                                                           6.06
                                                                     67.55 ( 3.08x)
25600
                  40
                                 400
                                         b
                                                  12
```

- Can see which activities account for most time
- Can see which activities limit parallel speedup

Some Logos



GraphChi: Going small with GraphLab



