

# Simulating Disaster Scenarios and Geographically-Correlated Resilient Overlay Networks

## Heuristics for Location-based Routing

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# Challenges

- ▶ **Goal:** Improve data delivery ratio/speed during geographically-correlated network failures
- ▶ No control over and limited knowledge about routing infrastructure
- ▶ But can gather location information
- ▶ **Claim:** Resilient Overlay Networks can help establish more reliable connections between end-hosts with limited topology knowledge

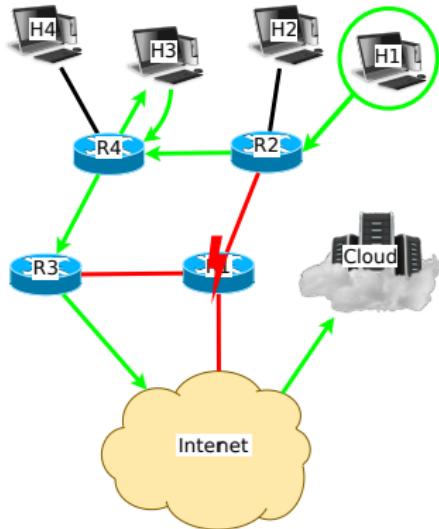
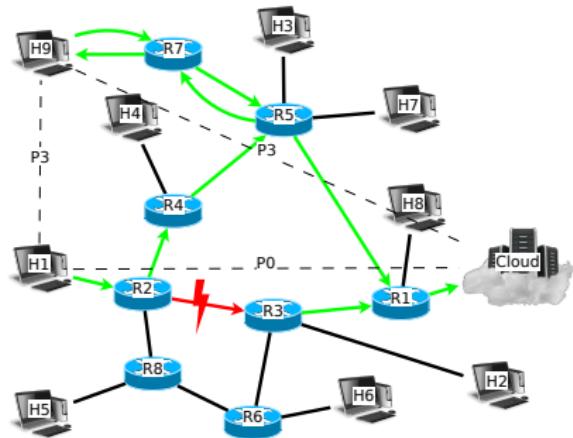


Figure: H1 routes around failed R1 using H3 as an overlay.

# Exploiting Location Information

- ▶ Contact overlay nodes outside of local region to avoid failures
- ▶ Choose overlay nodes from diverse regions not previously attempted
- ▶ Failure likely along path to server or in local area
- ▶ Choose node outside local region to avoid overlapping paths
- ▶ Choose path avoiding as much of the direct path as possible
- ▶ Overlay node may use similar route to sensor



**Figure:** Sensors try to contact nodes outside of the local area and not along the straight-line path to the server.

# Orthogonal Distant Path Heuristic

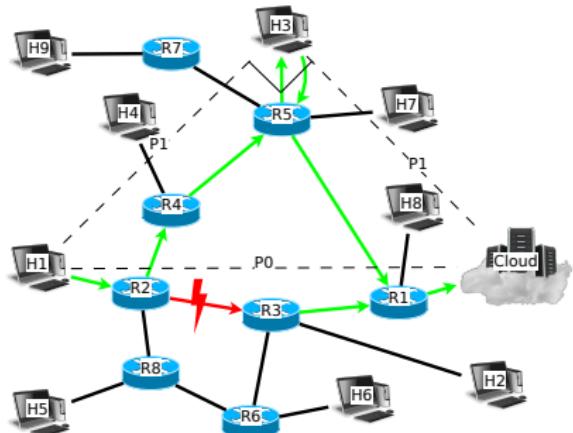
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## Algorithm 1:

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```
begin
    idealDist = |0.5 · dist(a, b)|
    perpDist = |sin(angA) · dist(a, c)|
    if angA >  $\frac{\pi}{2}$  or angA + angC <  $\frac{\pi}{2}$  or
    perpDist > dist(a, b) then
        likelihood = 0
    else
        likelihood = 0.5 · ((1.0 - (||perpDist| -
        |idealDist||/idealDist)2) + (1.0 -
        (| $\frac{\pi}{2}$  - angC|/ $\frac{\pi}{2}$ )2))
    end
```

---



**Figure:** The intuition of this heuristic is to avoid the straight path, without diverging from it too much. It strikes a middle ground by choosing a path at an ideal angle of  $45^\circ$ , which makes the angle at the top orthogonal, hence the name.

# New Region Heuristic

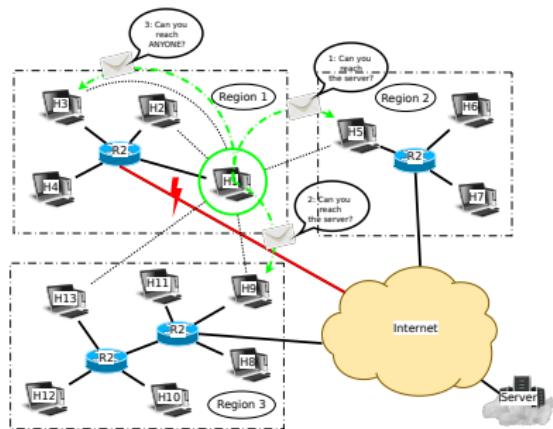
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## Algorithm 2:

---

```
begin
    if RegionsAttemptCount(region_count)
    then likelihood = 0
    else likelihood = 1.0
end
```

---



**Figure:** The intuition of this heuristic is to avoid regions that have been previously attempted unsuccessfully. We assume that no further attempts to contact such a region will succeed.

# New Angle Path Heuristic

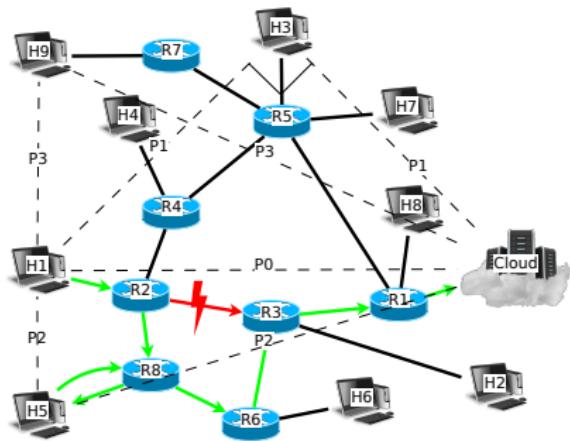
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## Algorithm 3:

---

```
begin
    if angle =  $\pi$  or angle = 0 then
        likelihood = 0
    else if angle <  $\pi$  then
        likelihood =  $\cos(\text{angle} - \frac{\pi}{4})$ 
    else if angle >  $\pi$  then likelihood =
         $\cos(2 \cdot ((2\pi - \text{angle}) - \frac{\pi}{4})/3)$ 
    likelihood *= pastLikelihood;
    /* aggregate likelihoods to make
    punishments on wrong choices */
end
```

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**Figure:** This heuristic attempts paths along new angles different from the ones previously attempted.

# Distance-Dependent Path Heuristic

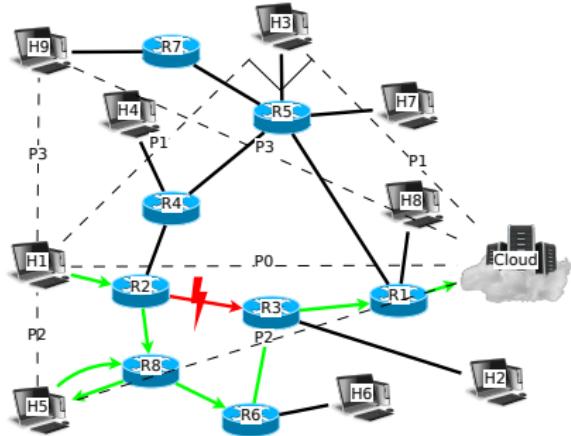
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## Algorithm 4:

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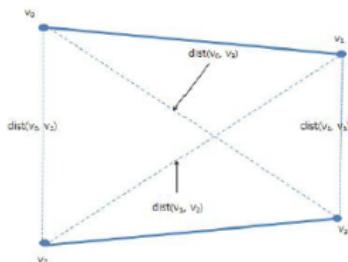
```
begin
    if dist <= minDist then
        likelihood = 0
    else if minDist < dist < idealDist
    then likelihood = dist2 / idealDist2
    else if dist >= idealDist then
        likelihood = idealDist / dist
end
```

---



**Figure:** This heuristic attempts paths that use overlay nodes at an ideal distance from the sensor. This ideal distance is chosen as a radius that reaches half-way to the server.

# Furthest First Path Heuristic

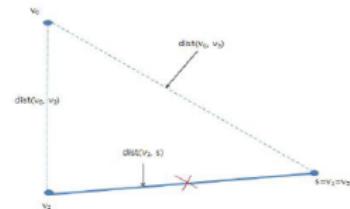


**Figure:** Definition of path similarity presented in W. Sun, “A Method for Overlay Network Latency Estimation from Previous Observation”, ICN-2013.

$$\text{geo\_similarity}(v_0, v_1, v_2, v_3) = \min[\text{dist}(v_0, v_2) + \text{dist}(v_1, v_3), \text{dist}(v_0, v_3) + \text{dist}(v_1, v_2)]$$

When considering one server,

$$\text{geo\_similarity}(v_0, v_1, v_2, v_3) = \text{dist}(v_0, v_2) = \text{geo}(v_0, v_2)$$



**Figure:** Thus, this heuristic considers overlay peers located further away to be more likely candidates.

---

**begin**

```
    /* Set maxDistance to a  
       considerably large number */  
    if  $\text{dist}(a, b) > \text{maxDistance}$  then  
        likelihood = 0  
    else likelihood = ( $\text{maxDistance} - \text{dist}(a, b)$ ) /  $\text{maxDistance}$ 
```

**end**

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# Closest First Path Heuristic

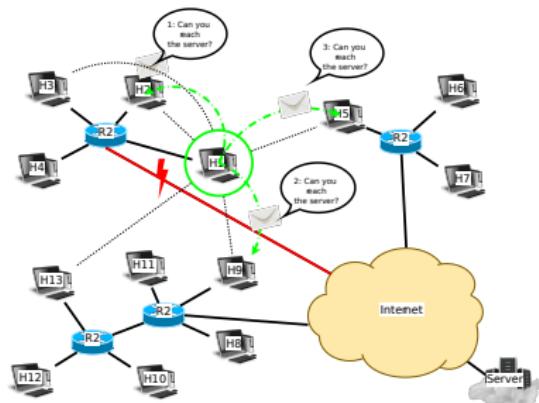
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## Algorithm 5:

---

```
begin
    /* Set minDistance to a
       considerably small number      */
    if  $dist(a, b) < minDistance$  then
        likelihood = 0
    else likelihood =
         $(dist(a, b) - minDistance)/dist(a, b)$ 
end
```

---



**Figure:** The intuition of this heuristic is to contact nearby overlay nodes that have found a path out of the local region. We found, however, that this approach does not work well in practice, likely due to the path similarity inherent with nearby nodes.

## Furthest First Path Heuristic

- ▶ Compared heuristics with baseline Random
- ▶ ns-3 simulation environment
- ▶ 10 seconds simulation time
  - ▶  $>10\text{s} \rightarrow$  data not as useful for early warning
  - ▶ Most damage done (earthquake over)
  - ▶ Routes start to recover
- ▶ Nodes retry every *timeout* seconds (up to 20 times)
- ▶  $\text{timeout} = 500\text{ms}$ : East  $\rightarrow$  West RTT  $\approx 200\text{-}300\text{ms}$
- ▶ Nodes/links in disaster region fail uniformly with probability  $p = 0.1, 0.2, 0.5$
- ▶ All overlay nodes in region (one incident link) report to server during disaster
- ▶ Server chosen randomly from 10 highest-degree nodes outside disaster
- ▶ Each combination of parameters repeated 100 times with different RNG seeds
- ▶ BRITE topology generator (10000 nodes, 50 AS'es, top-down Barabasi/Waxman model)