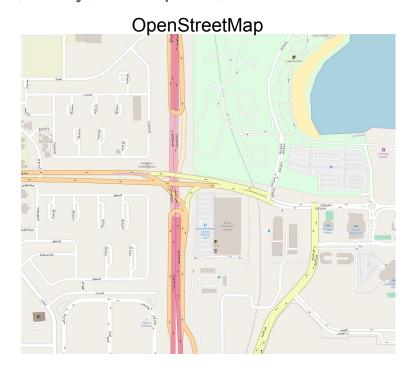
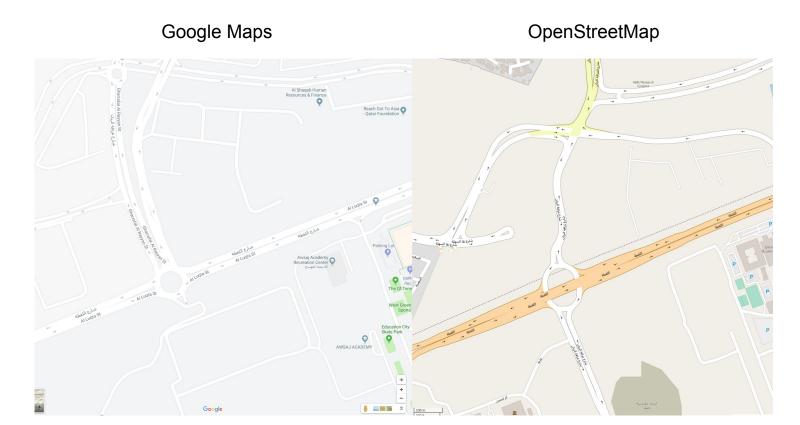
# Quality of Maps

- Examples
- Deepglobe challenge @CVPR by Facebook /SpaceNet
  - Complaints about ground truth
  - Provide example (Favyen/Songtao)
- SpaceNet Challenge

OSM is good for large scale implementations/comparisons But, noisy, incomplete, and erroneous in some areas.

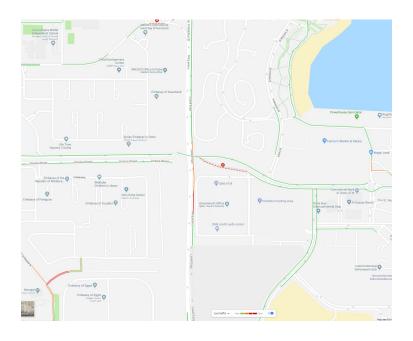






Even Google is not that accurate!





- Few more examples
- Should one use proprietary maps?
  - O Cost?
  - O Availability?
  - Scalability?

### Challenge 2: Getting the right data

#### **Satellite images?**









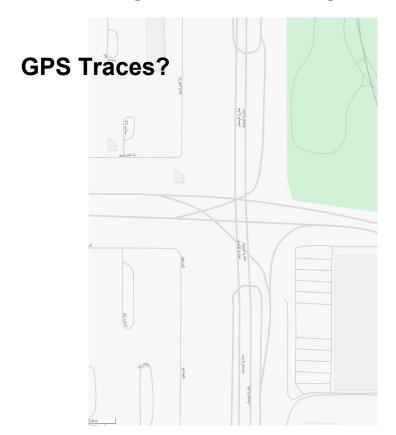
Google

Bing

DigitalGlobe Standard

DigitalGlobe Premium

### Challenge 2: Getting the right data





### Challenge 2: Getting the right data

Add some examples, from Doha changes, youtube video, etc.

### Challenge 3: Define the right evaluation Metric

#### Aspects to consider

- Structure and Geometry
- Function and Topology and Routing
- Humans vs. Machines:
  - o Beauty, smoothness, etc.

### Three Main Categories

- Pixel based distances
  - Precision/recall/F1
- Point set-based distances
  - Hausdorff
  - Holes and Marbles
- Path based distance
  - Path Length APLS (SpaceNet Challenge)
  - o Path Similarity Frechet.
- Intersection over Union?

## Pixel based distance

#### Pixel based distance

- Prominent in computer vision community
  - o Prediction, recall, F1
  - Relaxed precision, recall, F1
- Adopted in DeepGlobe 2018 @CVPR

#### Pixel based distance: Problem

- Ignores topology
  - Connections, intersections, under/over passes, etc.
  - Not good for routing

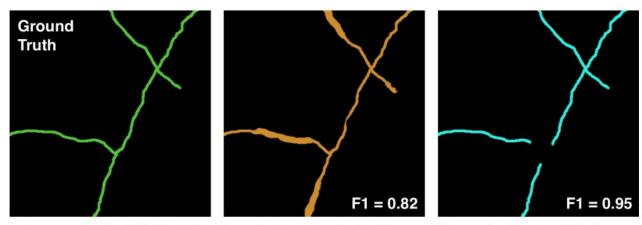


Figure 3. Example of the pitfalls of the pixel-based F1 metric. Left: Ground truth road mask in green. Middle: proposal mask in orange, which achieves a decent F1 score of 0.82 since most pixels are correctly labeled despite some inconsistencies in road width. Right: proposal mask in cyan, which achieves a superior F1 score of 0.95 since fewer pixels are classified incorrectly. For routing purposes, however, the rightmost plot is clearly inferior since it misses an important intersection and severs a road. The fact that pixel-based F1 scores incentivize the rightmost plot over the middle plot is suboptimal.

#### Credit. SpaceNet Challenge.

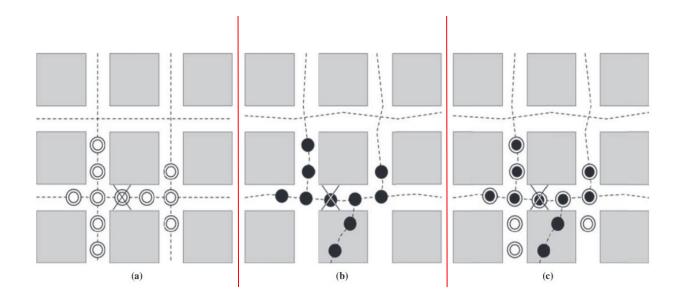
# Point set-based distance

#### Point set-based distances

- Each map is represented as a set of points
- Compute distance between points
- Hausdorff distance
  - No Topology.
  - Given two sets A, B;  $d(A,B) = max_a min_b(d(a, b))$
  - Very sensitive to outliers!
- Holes and Marbles
  - Sample the two maps
  - TOPO and GEO

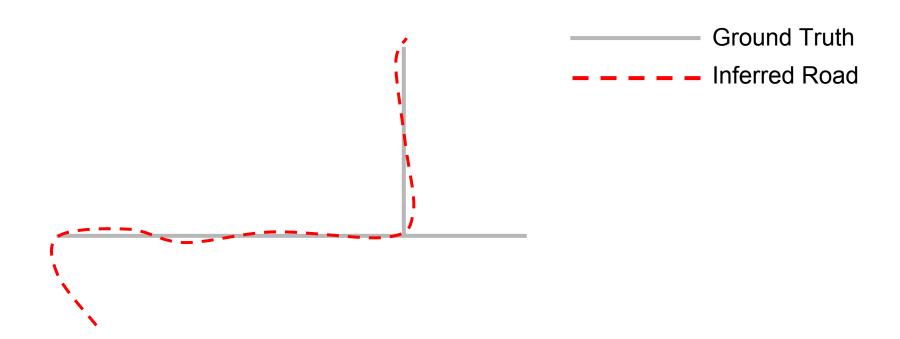
#### Holes and Marbles

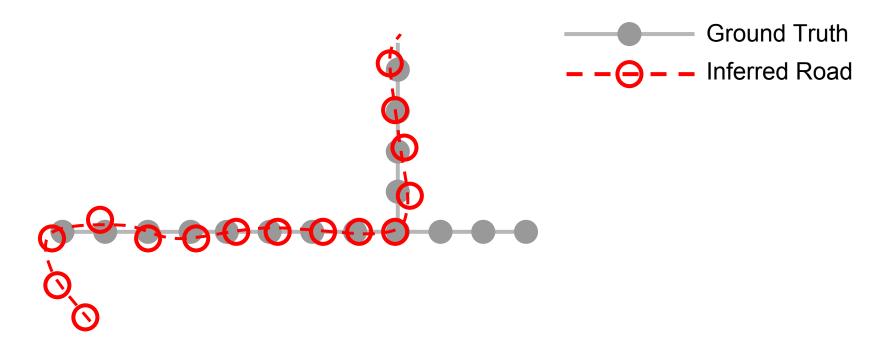
- GEO
- TOPO



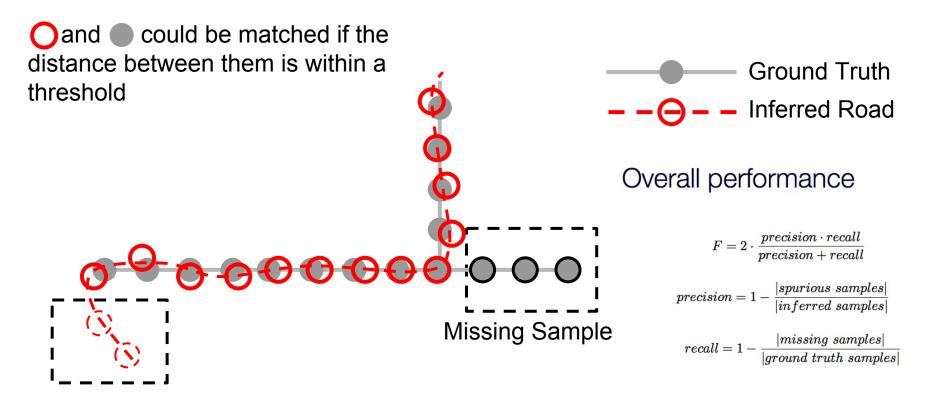
#### **Evaluation Metrics**

- Geometric evaluation (GEO)
- Graph-Sampling Based Distance (TOPO)
- Shortest Path Based Distance



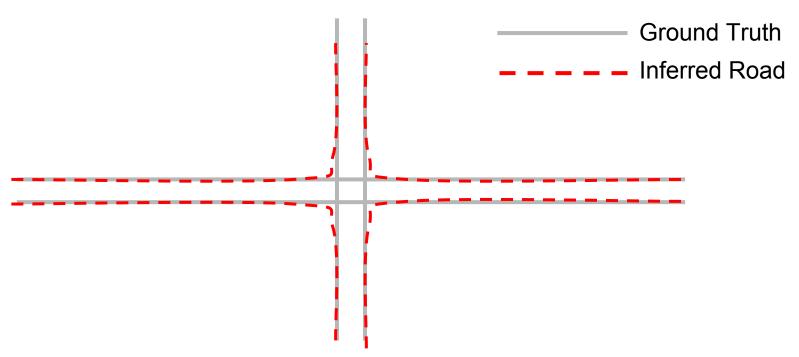


could be matched if the distance between them is within a **Ground Truth** threshold Inferred Road



**Spurious Samples** 

### Geometric evaluation (GEO) Limitation



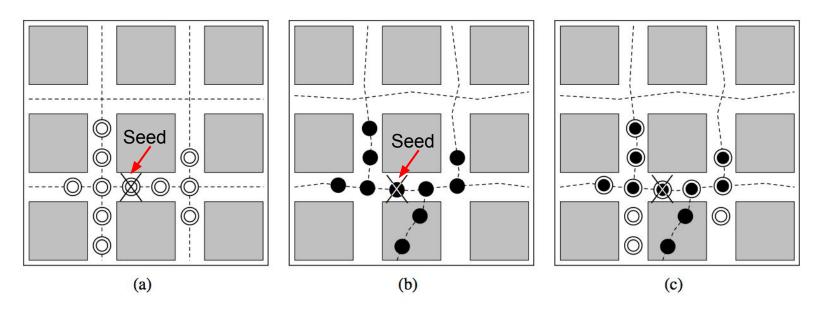
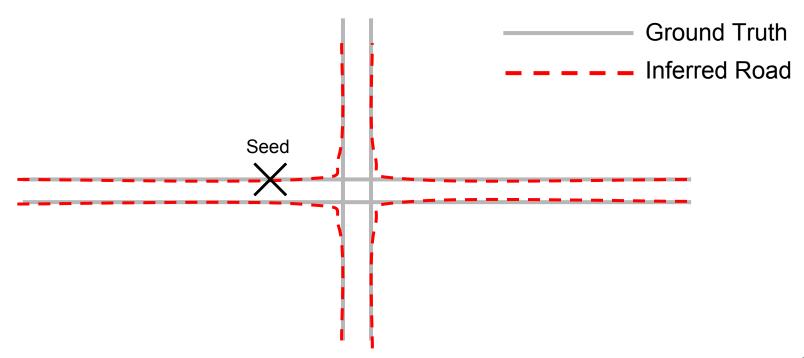
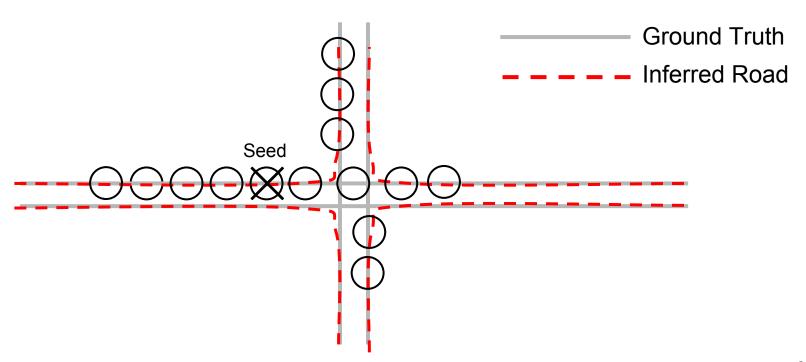
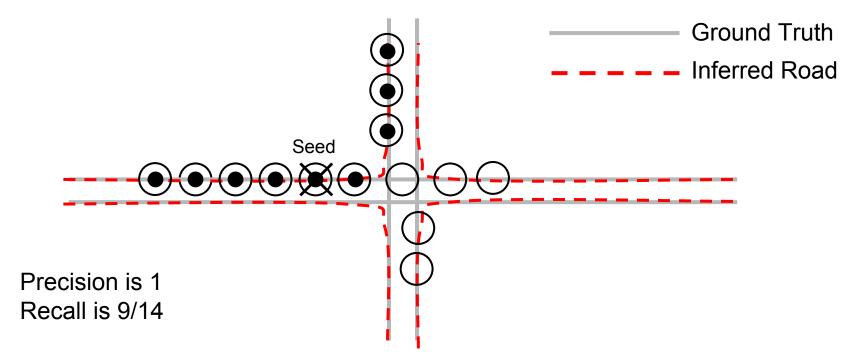


FIGURE 4 Overview of map comparison algorithm. (a) Holes are dropped at even intervals along edges of the ground truth map. (b) Marbles are dropped at even intervals along edges of the generated map. (c) Marbles from generated map fill holes where the maps overlap.

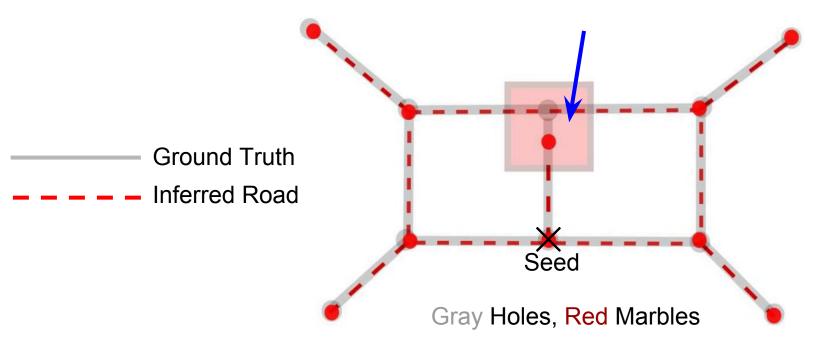






### Graph-Sampling Based Distance (TOPO) Limitation

TOPO may fail to capture the broken connection

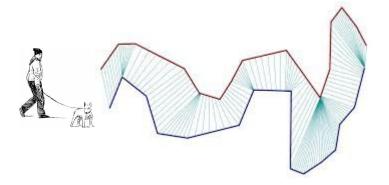


Ahmed, Mahmuda, et al. "A comparison and evaluation of map construction algorithms using vehicle tracking data." GeoInformatica 19.3 (2015): 601-632. 28

# Path based distance

#### Path based distances

- Each map is represented by a set of paths.
- Unique paths, shortest paths, sampling
- Path similarity
  - Discrete frechet distance.
  - Frechet distance idea: the minimum leash length such that a man and a dog can walk on the two curves (paths) generated from G.
  - o  $d(P_A, P_B) = min_{a \in PA} max_{b \in PB} (Frechet(a, b))$
- Path length
  - APLS



### APLS: Average Path Length Similarity

- Graph theory based
- Considers topology
- Good for routing

$$C = 1 - \frac{1}{N} \sum min \left\{ 1, \frac{|L(a,b) - L(a',b')|}{L(a,b)} \right\}$$

$$N = number of unique paths$$
  
 $L(a,b) = length of path(a,b)$ 

Equation 1. APLS metric. Node a' is the node in the proposal graph G' nearest the location of node a in the ground truth graph G. L(a,b) denotes a path distance in the ground truth graph G, and L(a',b') denotes the path length between the corresponding nodes in the proposal graph.

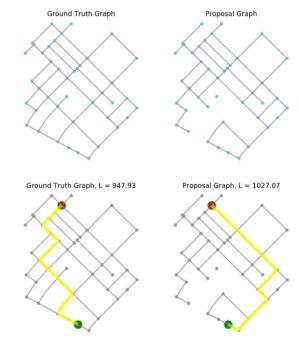


Figure 4. Demonstration of path length difference between sample ground truth and proposal graphs. Upper Left: Ground truth graph. Upper Right: Proposal graph with 30 edges removed. Lower Left: Shortest path between source (green) and target (red) node in the ground truth graph is shown in yellow, with a path length of –948 meters. Lower Right: Shortest path between source and target node in the proposal graph, with a path length of –1027 meters; this difference in length forms the basis for our graph similarity metric. Plotting is accomplished via the excellent osmnx package.

#### APLS vs. F1

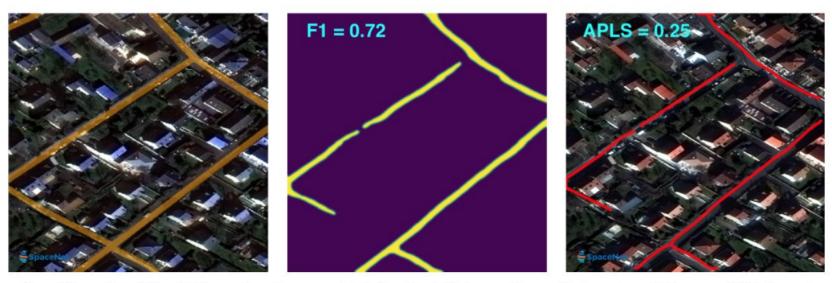


Figure 6. Comparison of F1 and APLS scores for a given proposal. Left: Sample satellite image with ground truth mask overlaid in orange. Middle: Proposal mask in yellow, yielding an F1 score of 0.72. Right: Proposal graph (red) inferred from the proposal mask; missing intersections and road segments are heavily penalized, yielding an APLS score of 0.25.

#### **APLS**

How to map nodes in the GT to nodes in the generated map?

Node semantics: we define control nodes as road endpoints, intersections, or midpoints along routes of interest.

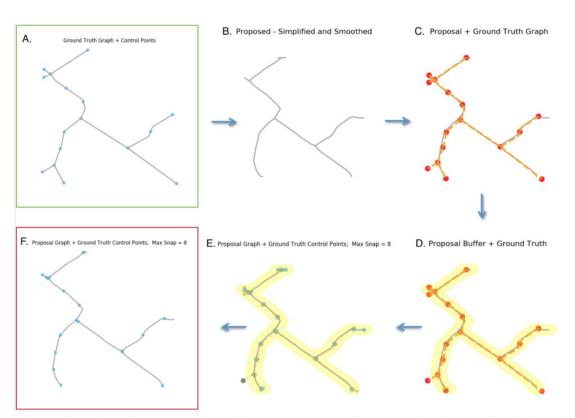


Figure 1. Node snapping procedure. A: Ground truth graph with control nodes. B: Proposal graph. C: Ground truth graph (orange) and ground truth control points (red) overlaid on proposal graph (grey). D: Same as graph C, but showing an 8 meter buffer (yellow) around the proposal graph. E: Ground truth control nodes are injected into the proposal graph at the nearest edge, except for nodes outside the buffer (grey). F: Final proposal graph with nodes injected at the appropriate location to compare to graph A.

### APLS: Symmetric comparison

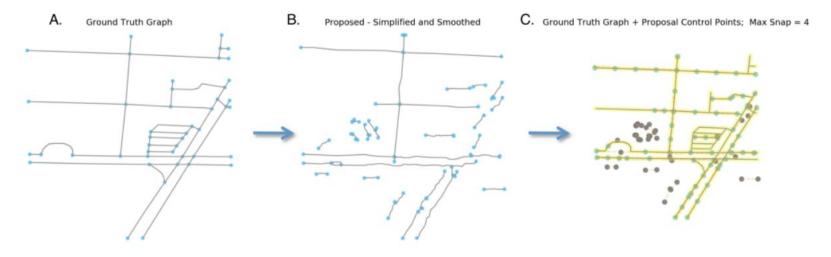
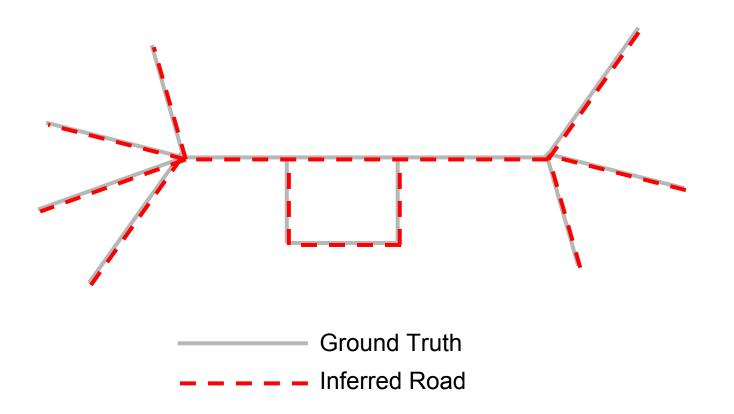
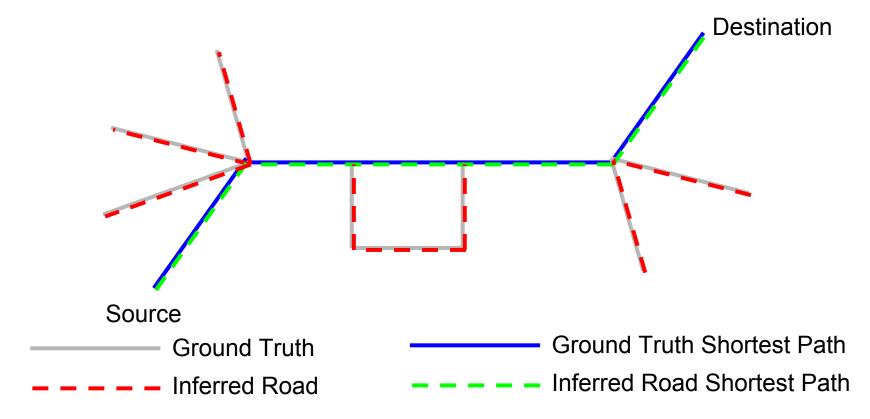
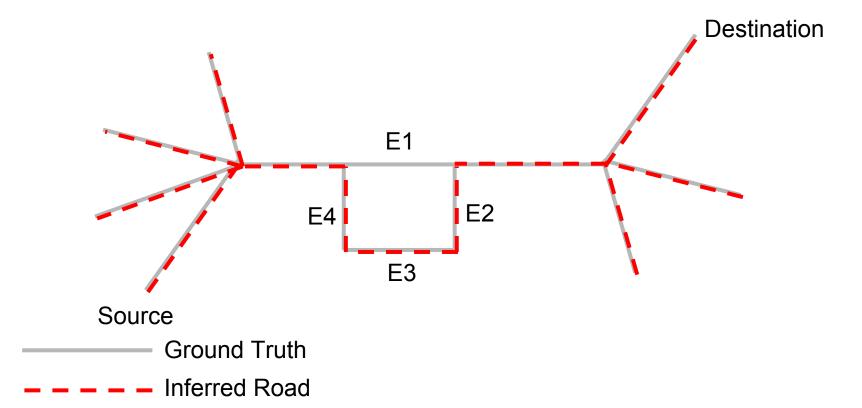
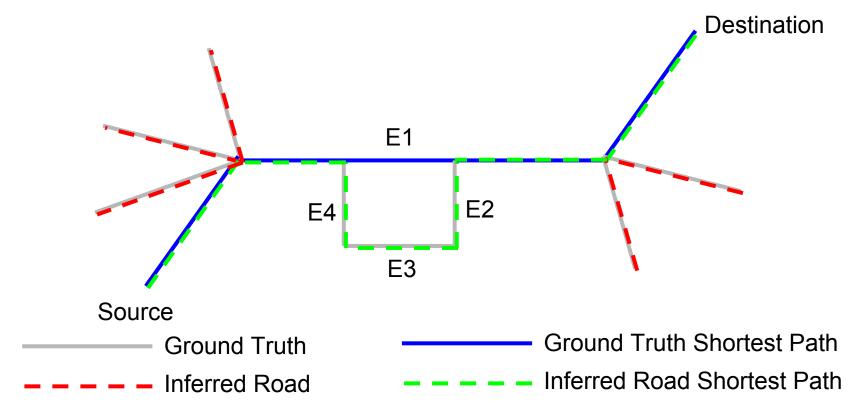


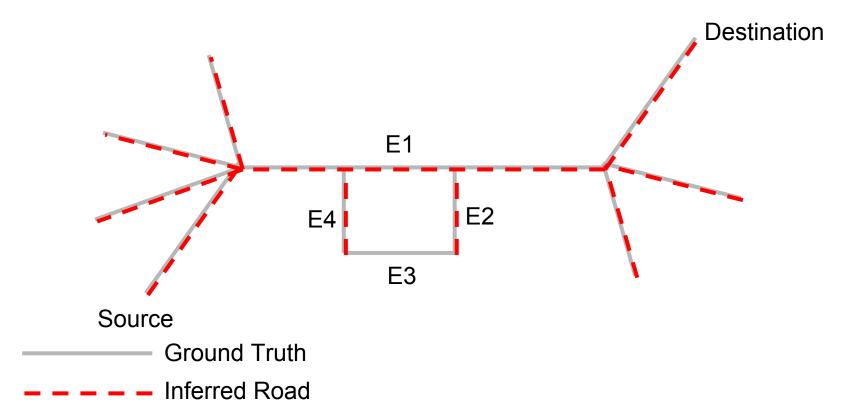
Figure 2. Illustration of the need to apply Equation 1 symmetrically (i.e.: ground truth to proposal, and proposal to ground truth). A: Ground truth graph. B: Proposal graph with many short, spurious connections well outside the buffer. C: Proposal nodes snapped onto the ground truth graph; snapping proposal control points onto the ground truth graph and then computing the metric penalizes the many extraneous predictions (grey nodes).

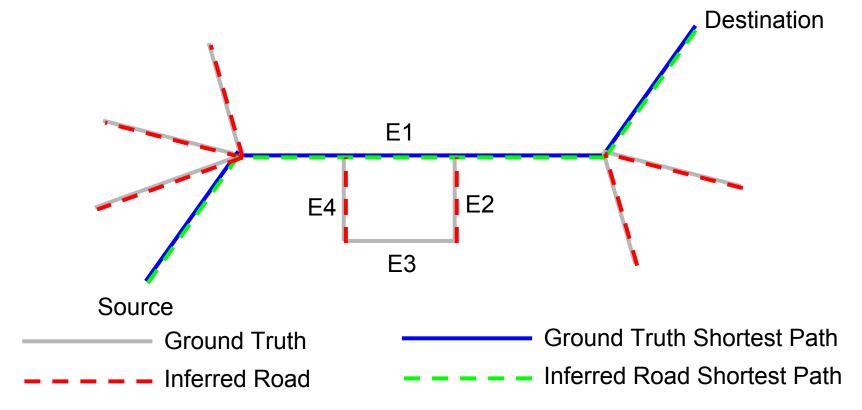


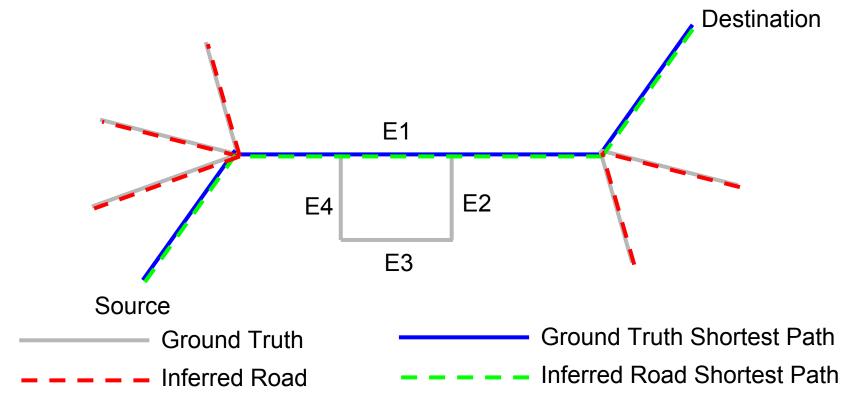












#### Literature review

Check all Map Making papers, especially recent ones, and summarize metrics they used.

#### References

- [relaxed Precision/Recall] Zhang, Z., Liu, Q., & Wang, Y. (2018). Road extraction by deep residual u-net. *IEEE Geoscience and Remote Sensing Letters*.
- [relaxed Precision/Recall] Mnih and G. Hinton, "Learning to detect roads in high-resolution aerial images," ECCV, pp. 210–223, 2010

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#### Methods from Ahmed et al.

### Challenge 3: Define the right evaluation Metric

- Directed Hausdorff distance
- Path based distance
- Shortest paths based distance
- Holes and marbles
- Intersection over Union
- Precision/recall/F1, relaxed precision/recall (pixel within x labeled pixel.)