

# **Automatic Alignment of Vector Data and Orthoimagery for The National Map**

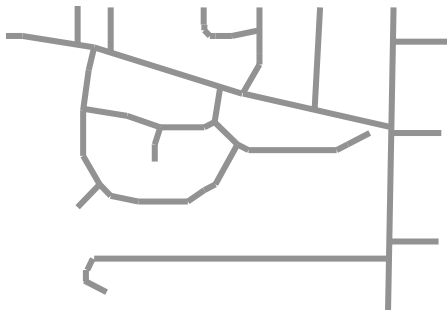
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# Outline

- Introduction & Motivation
  - The National Map
- Our Approach to align vector and imagery
  - Approach overview
  - Improvements over our previous approach
- Related Work
- Conclusion and Future Work

# Introduction

- Geospatial data sources have become widely available
- Automatically and accurately integrating and aligning two spatial datasets is a challenging problem



Road network  
( in vector format )



Orthoimagery  
( in raster format )



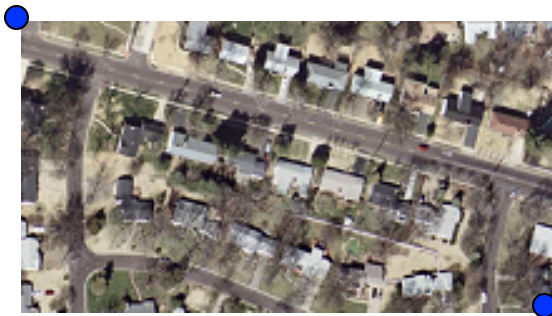
Street maps  
( in raster format )

# Motivation: Vector and Imagery Integration

## Challenges

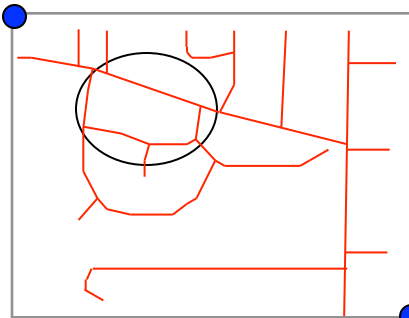
- Different projections, accuracy levels, resolutions result in spatial inconsistencies

Lat / Long



Lat / Long

Lat / Long



Lat / Long





# Motivation: The National Map

- The National Map is a government effort to make geospatial data available for 133 urban areas of the US for Homeland Security
- Purpose is to make these integrated datasets available to government organizations to support crisis response and emergency planning, etc.
- There are no automated techniques for aligning vector data with orthoimagery and this is a very labor intensive task.

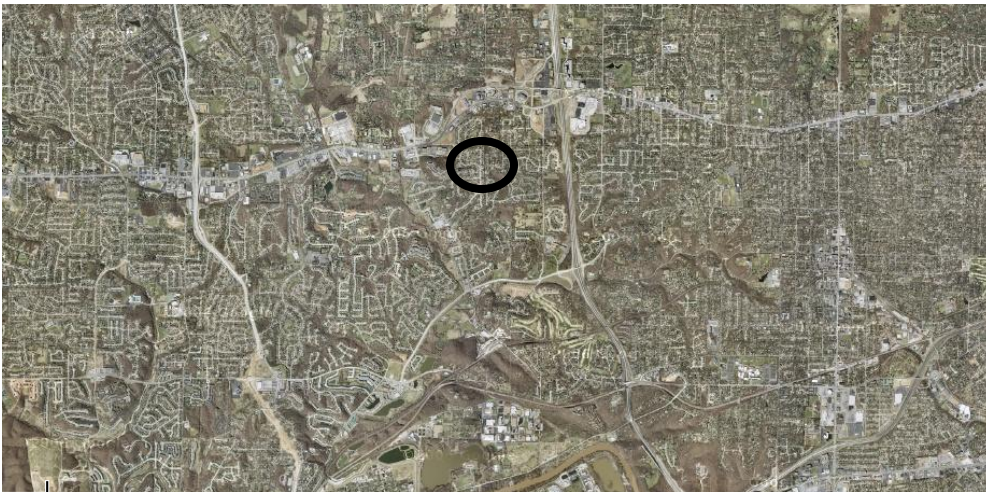
# Motivation: The state of the art

- Traditionally, the problems of vector-imagery and map-imagery alignment have been in the domain of GIS and Computer Vision
- In GIS literature
  - The alignments were previously performed manually
    - Commercial products: *ESEA MapMerger* ESRI ArcView; *Able R2V*; *Intergraph I/RASC*
- In Computer Vision literature
  - Alignment was performed automatically based on image processing techniques
    - Often required significant CPU time
    - Accuracy quite poor

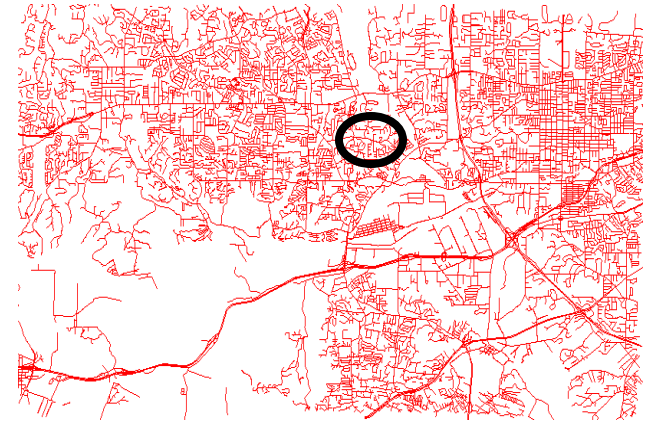
# The example

- The Data Sets (for the National Map)
  - USGS high resolution color imagery
  - Road vector data from DOT, MO

USGS 0.3m/p color imagery



Road network



**They are misaligned, and there is no global transformation**



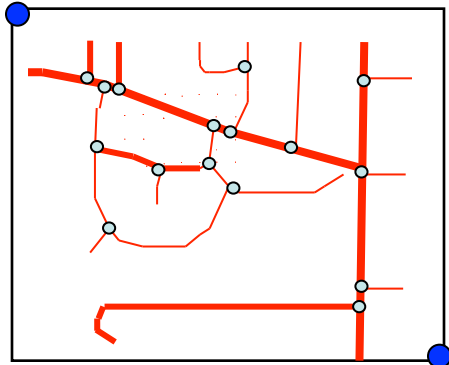
# The Vector-Imagery conflation approach

Lat / Long



Lat / Long

Lat / Long



Lat / Long

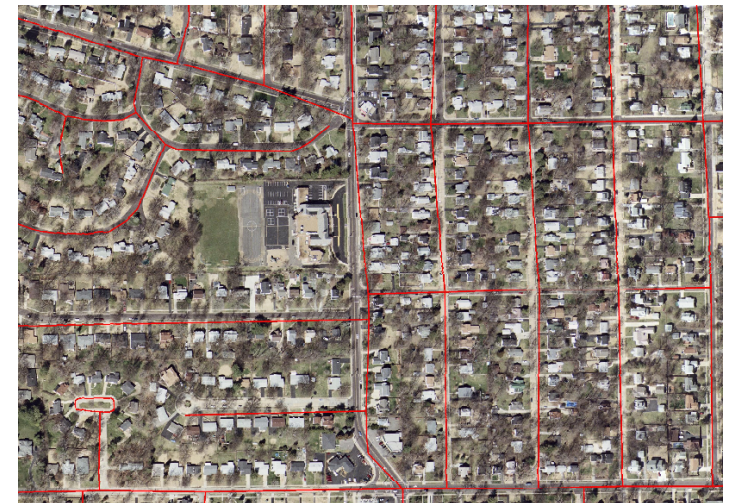
Control Point  
Detection

Filtering  
Technique

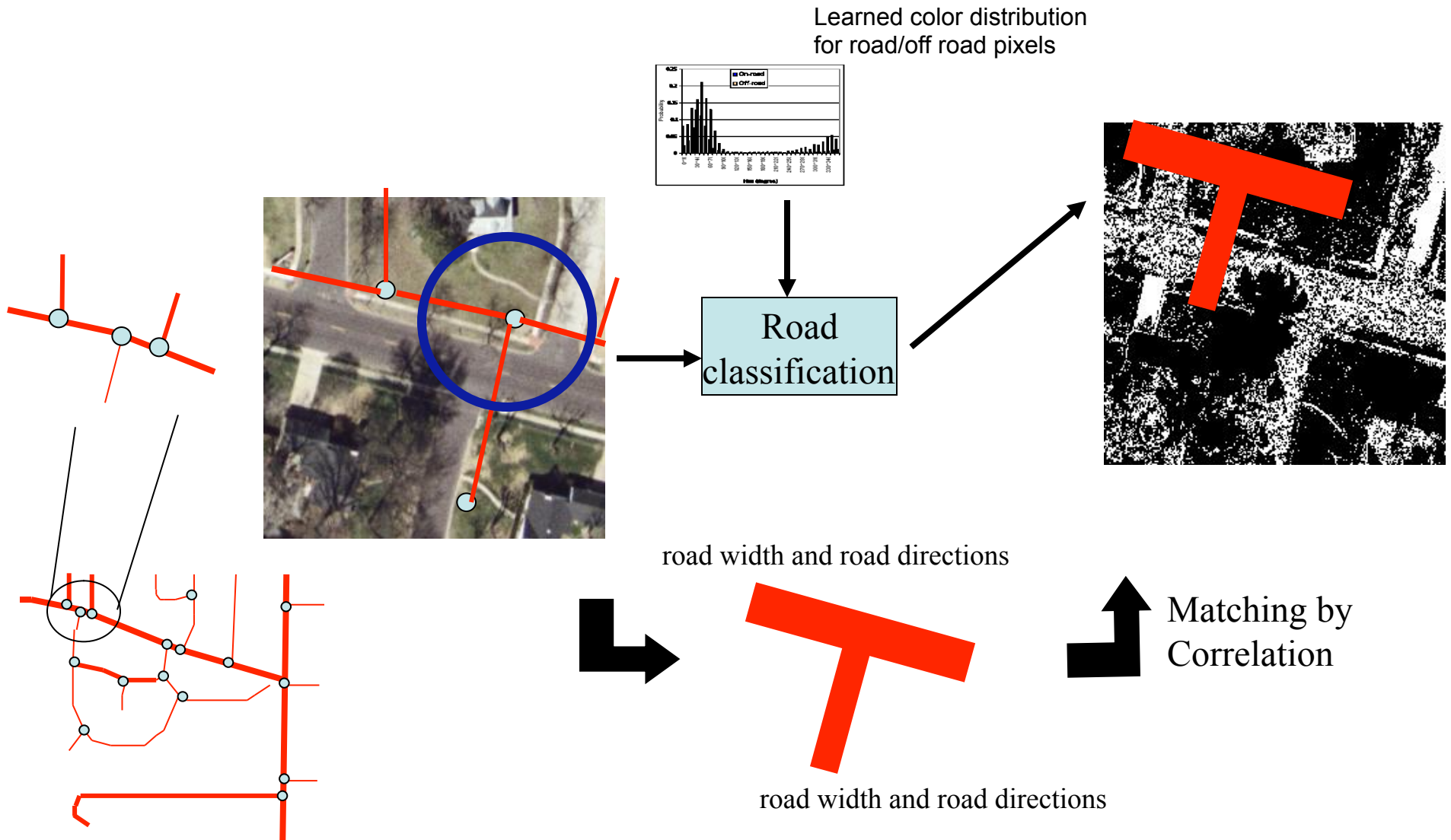
Intermediate  
control points

Final  
control points

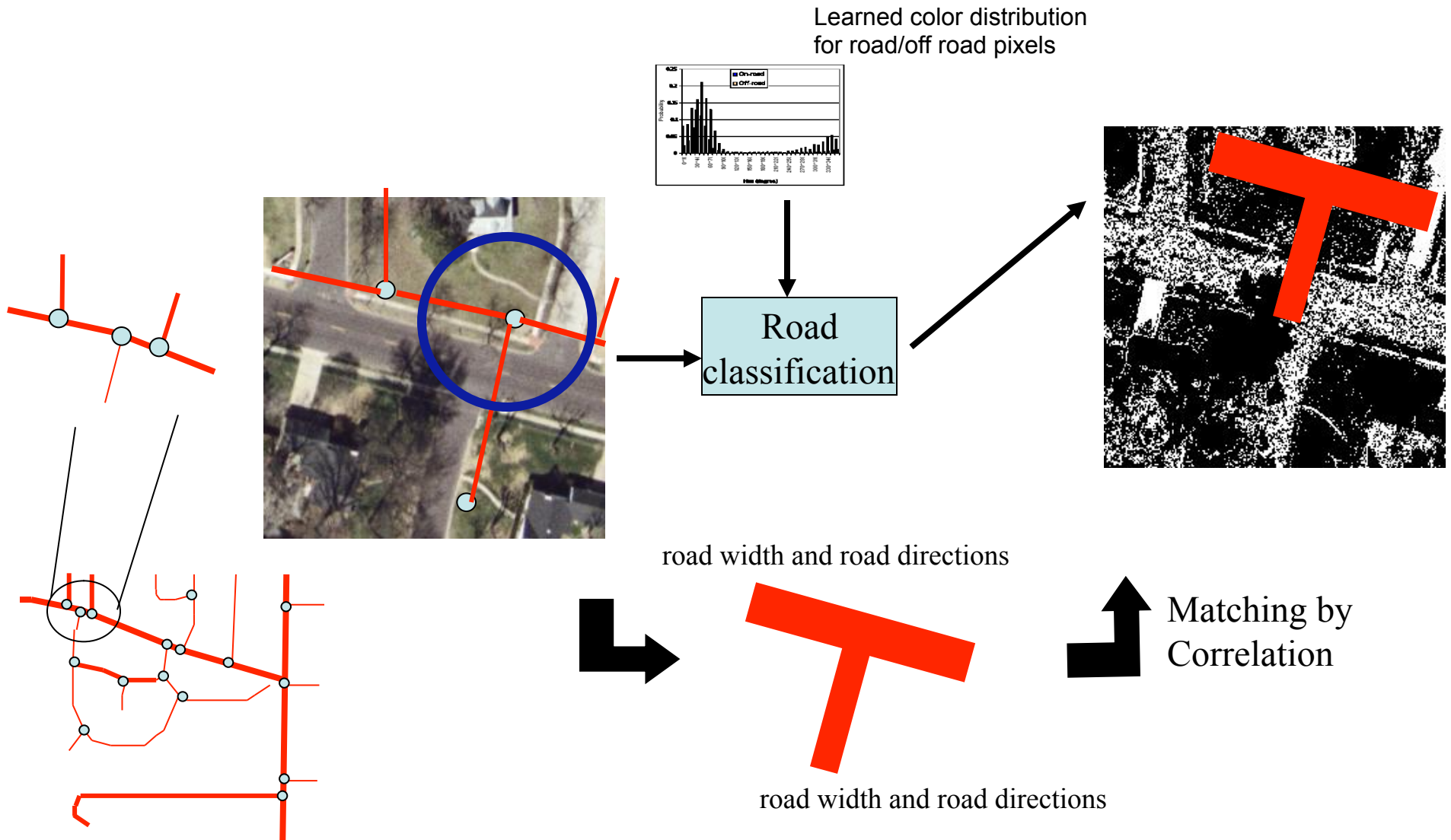
Triangulation and  
Rubber-Sheeting



# Finding Control Points Using Localized Template Matching

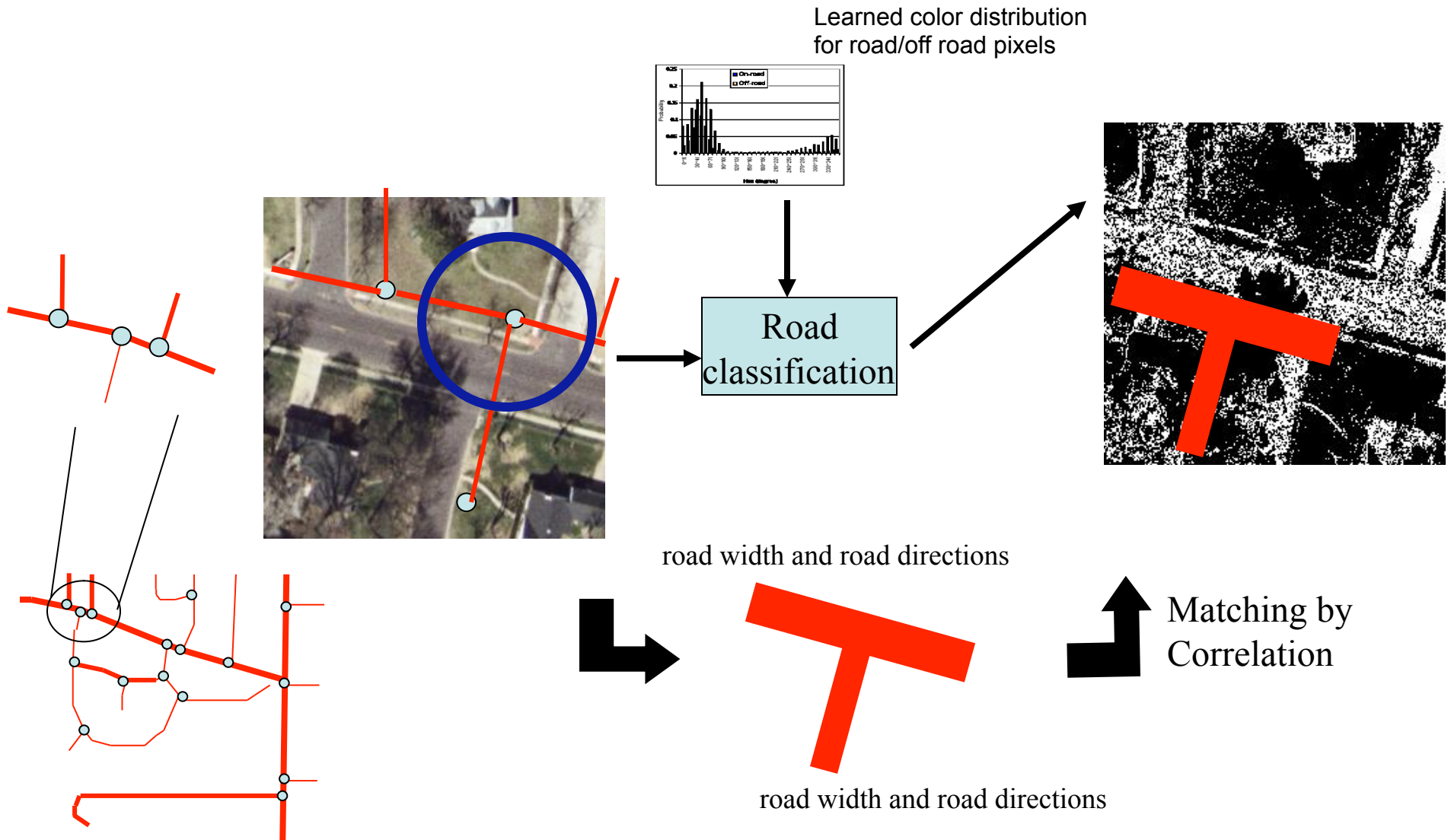


# Finding Control Points Using Localized Template Matching

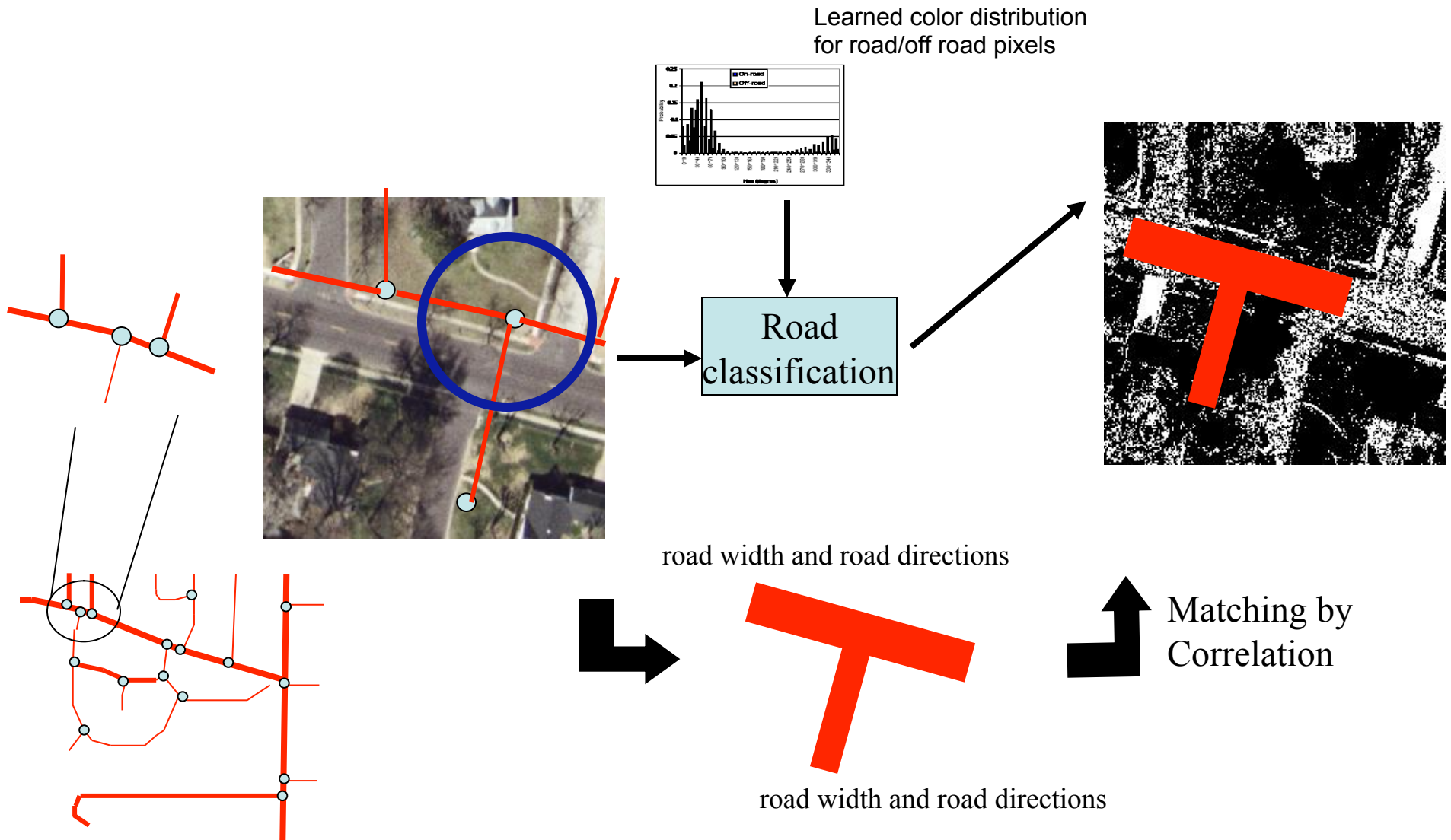




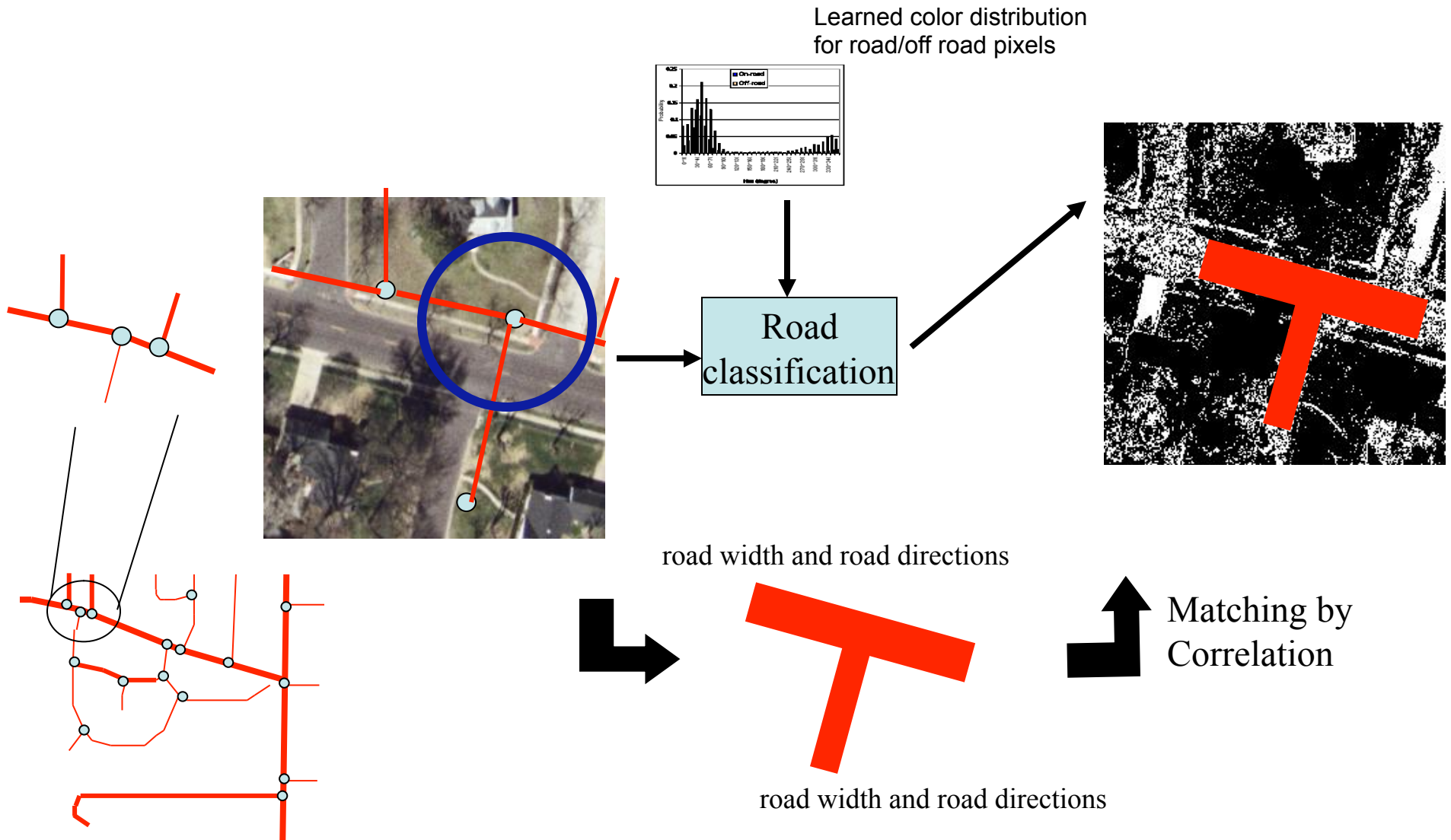
# Finding Control Points Using Localized Template Matching



# Finding Control Points Using Localized Template Matching

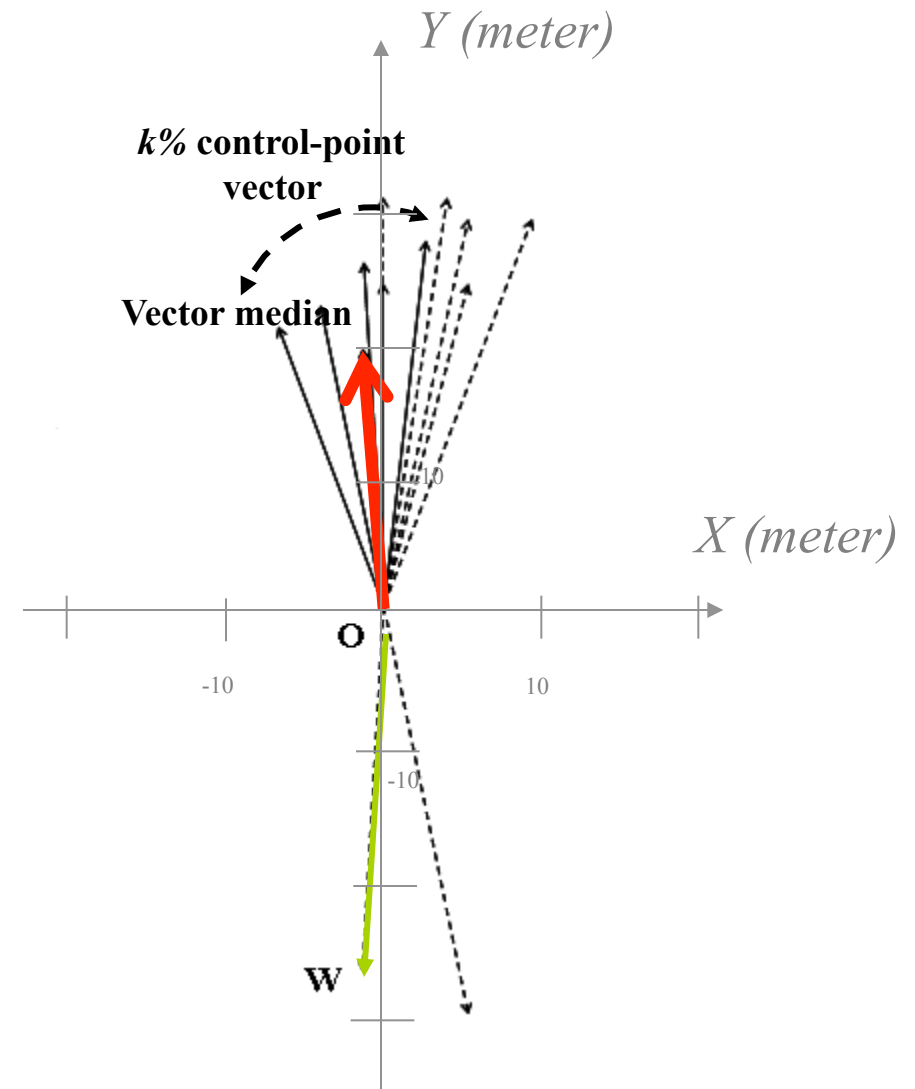


# Finding Control Points Using Localized Template Matching



# Filtering Control Points Using Vector Median Filter (VMF)

- View the control point pair displacement as vector
- Using a fixed ratio ( $k\%$ ) to keep control point pairs that have similar displacement as the median one

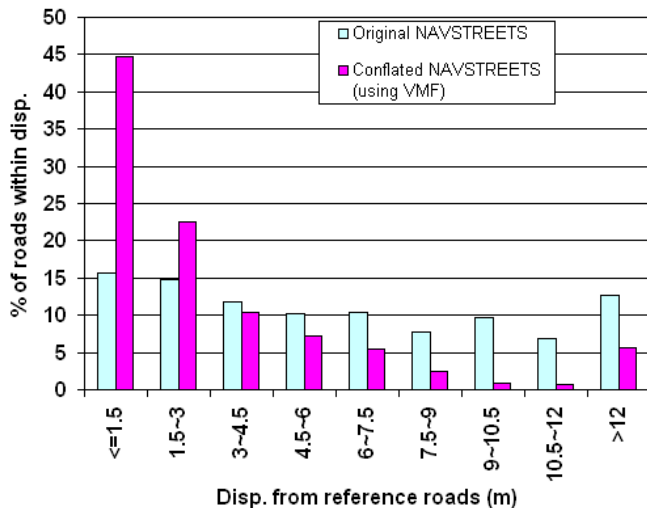




# Results:NAVSTREETS + High-res Image

	Original NAVSTREETS	Conflated NAVSTREETS
<b>Completeness</b>	<b>44.9 %</b>	<b>74.4 %</b>
<b>Correctness</b>	<b>47.9 %</b>	<b>85 %</b>

## Positional Accuracy

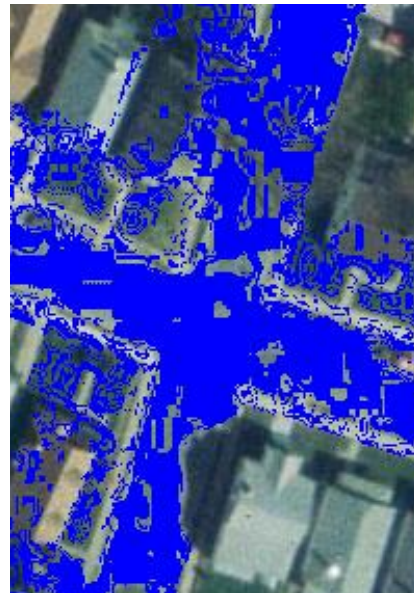


# Road Classification Used in Localized Template Matching

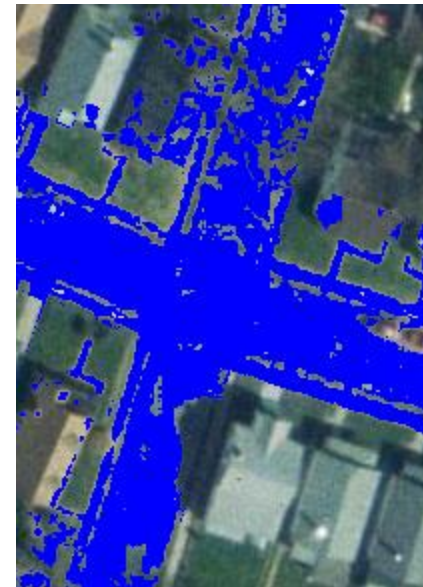
- Before: Bayes classifier based on Hue component of learned road/off-road pixels
- Improved: Support Vector Machine (SVM) classifier based on all color channels (R,G,B) of learned road/off-road pixels
  - Much fewer “false positives” and more “true positives”



Original imagery



Road-classified pixels  
based on Bayes classifier

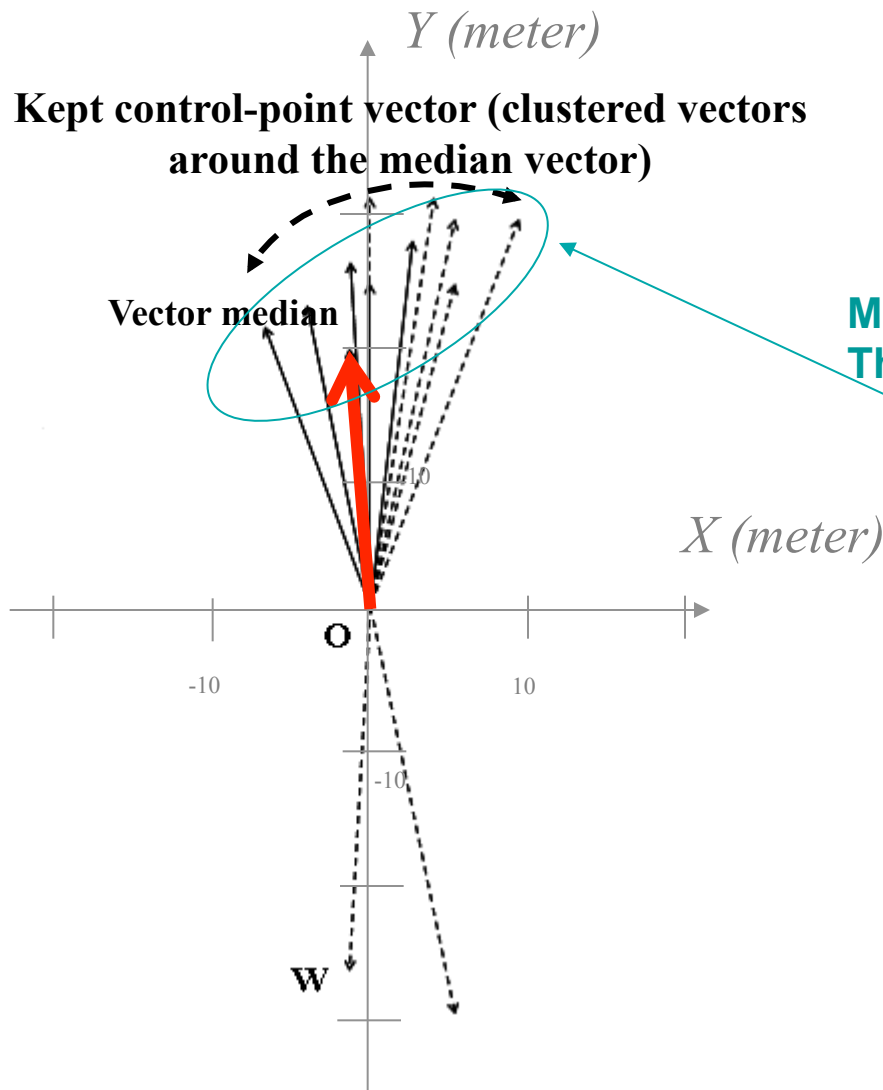


Road-classified pixels  
based on SVM classifier

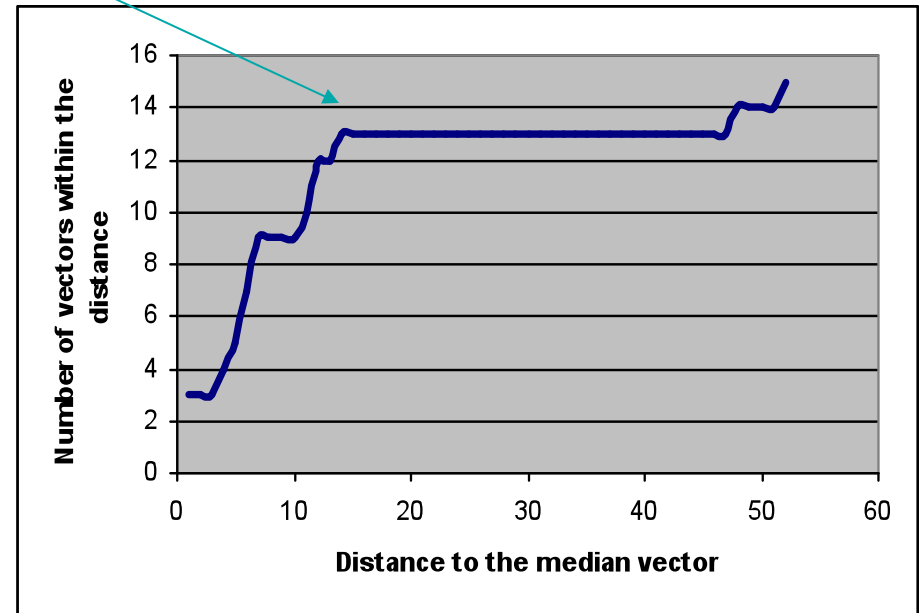


# Filtering Control Points Using Vector Median Filter (VMF)

- Improved: Dynamic determine the ratio
  - Investigate the cluster around the median vector
  - Accommodate more control point pairs



Most of the vectors are close to the median vector. This forms a cluster around the median vector



# The Vector-Imagery conflation approach: Triangulation and RubberSheeting





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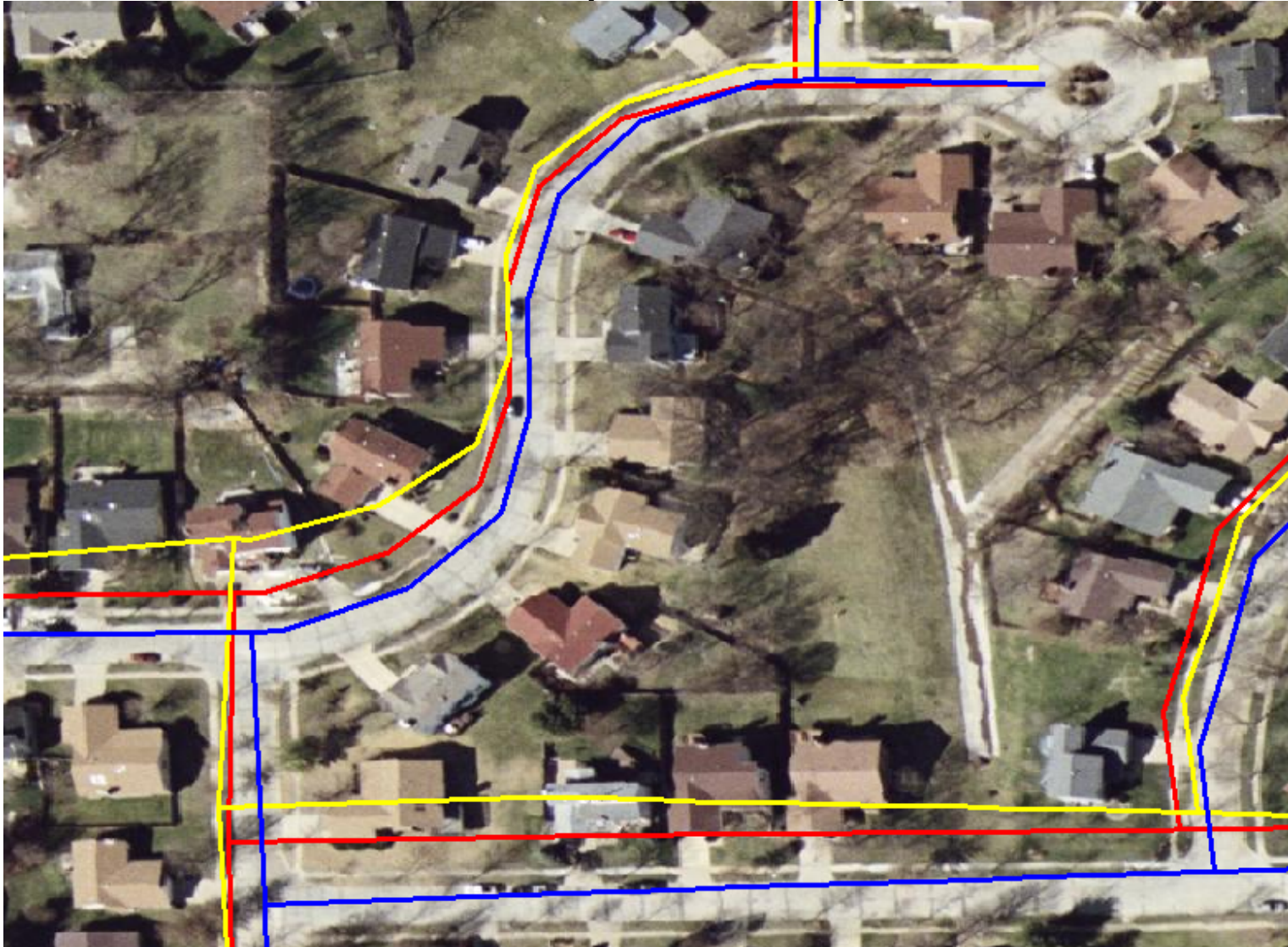


# Improved results: comparing with results based on previous technique

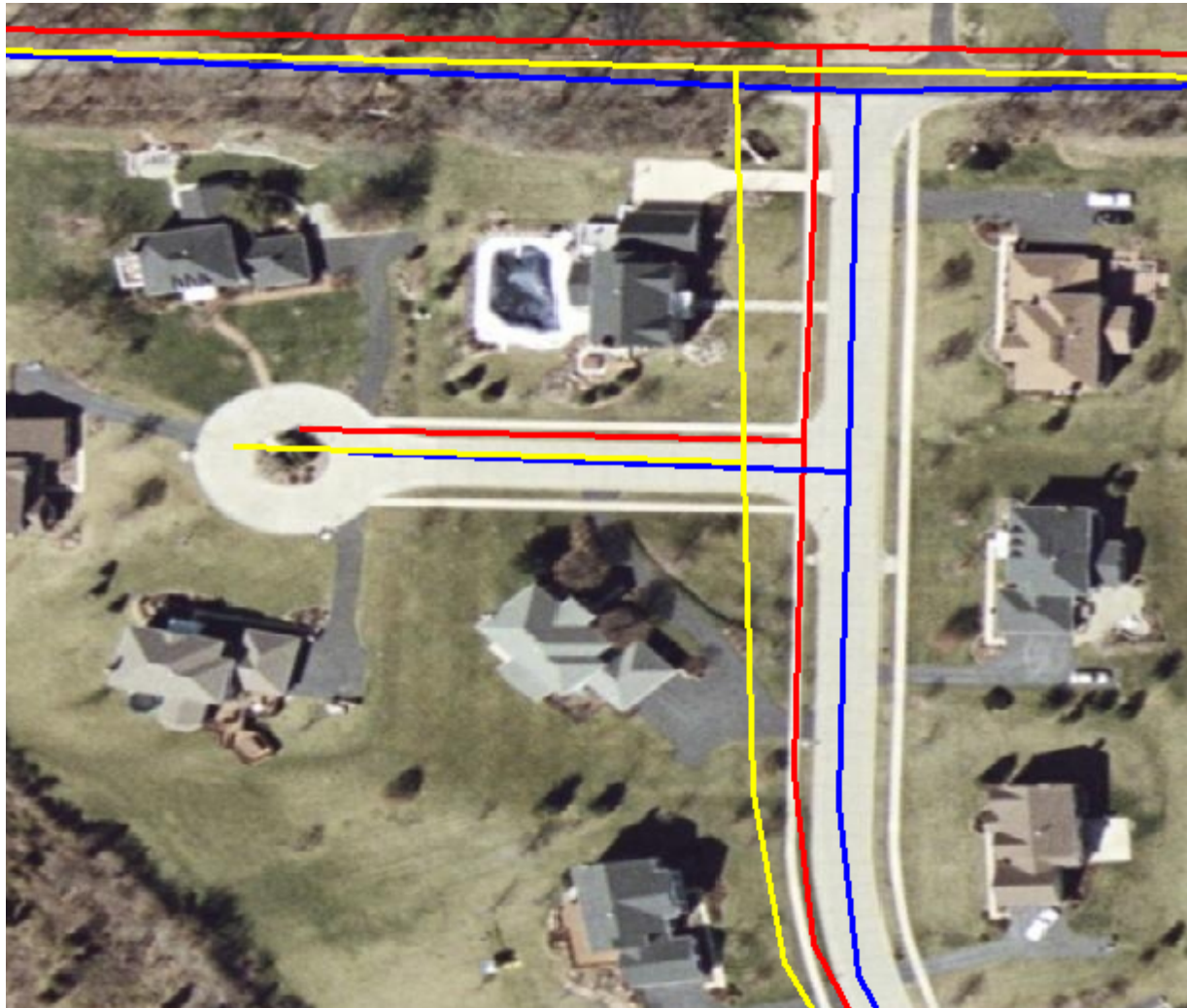
**Red lines:** Original roads

**Yellow lines:** Conflated results based on previous technique

**Blue lines:** Conflated results based on improved technique



More improved results: comparing with results based on previous technique





# Related Work

- Vector to imagery conflation
  - Utilizing matched polygons [Hild et al. 98]
  - Utilizing matched lines [Filin et al. 00]
  - Utilizing matched junction-points [Flavie et al. 00]
  - All above solutions
    - Require lots of CPU time
    - Utilize vector data only for verifying detected features not for extracting features
  - Commercial products: ESRI ArcView
    - Pick control points manually

# Conclusion

- Accomplishments
  - ✓ Refinement of pattern recognition procedures for identifying the road intersections in the images
  - ✓ Refinement of the filtering procedures for the ground control points
  - ✓ Development of methods for matching across image panels
  - ✓ Overall improvement of the accuracy of the transformed transportation data to match the images

# Future Work

- Address the alignment of road vector data with highways
- Apply the same techniques to automatically align:
  - Vector Parcel Data
  - Hydrographic Data
  - Elevation Data