

Understanding the Potential for Video Analytics to Support Traffic Management Functions



ENTERPRISE Program

Program Goals

- Facilitate rapid progress in the development and deployment of ITS technologies
- Accelerate the systematic advancement of selected ITS projects

Members carry out ITS projects and activities including fundamental research, technology development, demonstration, standardization, and deployment.

ENTERPRISE Program

Members

- Arizona DOT
- Georgia DOT
- Idaho Transportation Department
- Illinois DOT
- Iowa DOT
- Kansas DOT
- Maricopa County, AZ
- Michigan DOT
- Minnesota DOT
- Mississippi DOT
- Oklahoma DOT
- Pennsylvania DOT
- Texas DOT
- Washington State DOT
- Ontario Ministry of Transportation
- Transport Canada
- Dutch Ministry of Transport
- FHWA

What is Video Analytics?

Video Analytics systems **process video streams** from traffic cameras to:

- **Collect Traffic Data:** Vehicle counts, speeds, vehicle classifications
- **Detect Incidents and Create Alerts:** Stopped vehicles, slow traffic, wrong-way vehicles, wildlife, pedestrians, debris

DATE	TIME	VOL(NS)	SPEED(NS)
2013/06/13	12:00 AM	161	71
2013/06/13	12:00 AM	130	68
2013/06/13	12:15 AM	121	70
2013/06/13	12:30 AM	112	69
2013/06/13	12:45 AM	83	70
2013/06/13	1:00 AM	54	67
2013/06/13	1:15 AM	59	68
2013/06/13	1:30 AM	84	68
2013/06/13	1:45 AM	48	65
2013/06/13	2:00 AM	74	71
2013/06/13	2:15 AM	60	66
2013/06/13	2:30 AM	65	68
2013/06/13	2:45 AM	56	68

Traffic Data Output (15-min Increments)



Incident Detection

Why Use Video Analytics?

Challenges

- Difficult to monitor conditions in rural areas
- Challenge for TMC operators to monitor multiple camera views simultaneously
- Vehicles traveling the wrong way introduce safety hazard

Opportunities

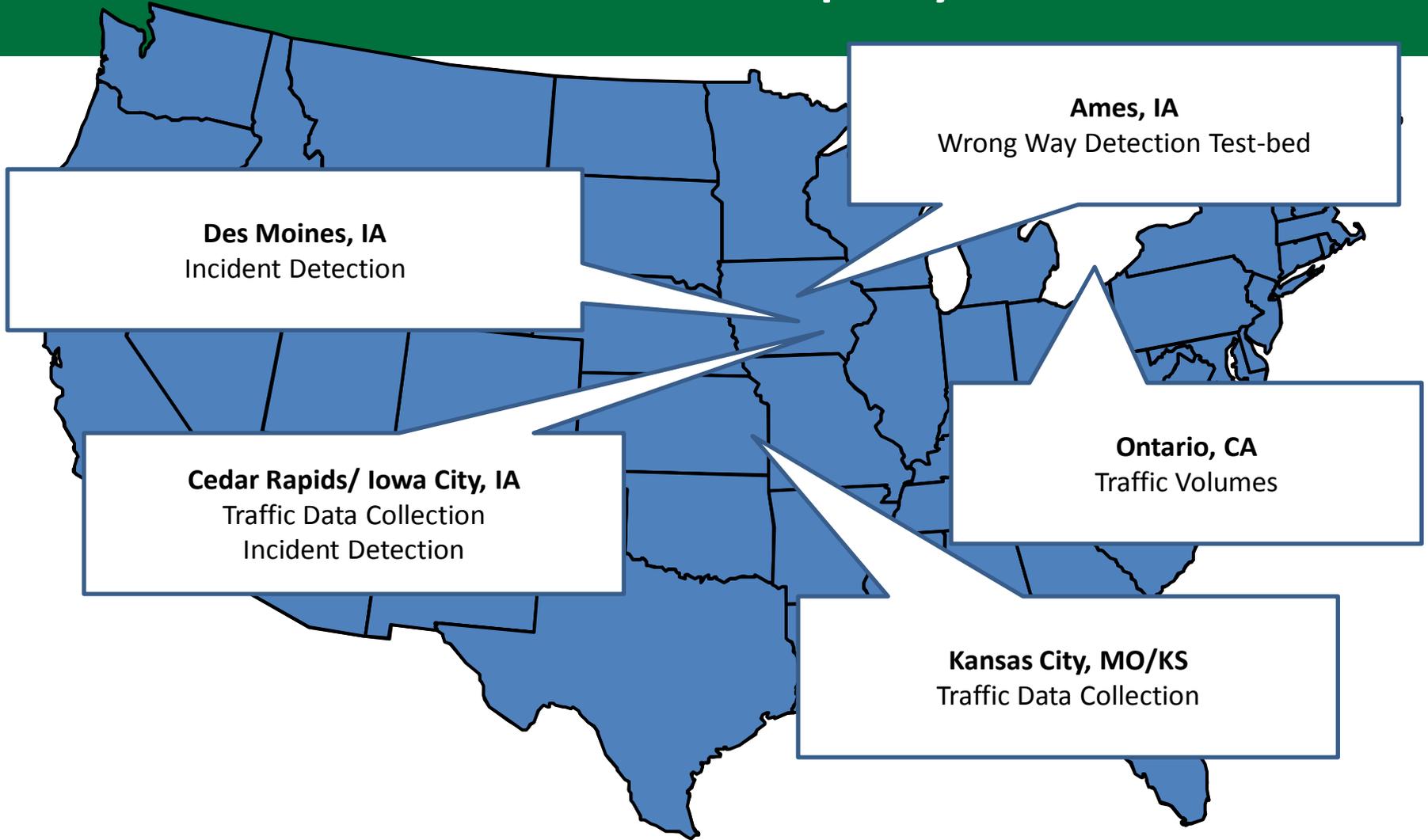
- Utilize existing camera infrastructure
- Potential to use Video Analytics for multiple purposes (traffic data collection, incident detection)

Why Evaluate Video Analytics?

Project Goals

- Investigated potential of Video Analytics as a tool for:
 - Traffic data collection
 - Incident detection
 - Wrong-way vehicle detection
- “Proof of Concept” evaluation to understand current state of practice
 - How accurate? How effective? How useful?
 - Compared to traditional methods/technologies: Loop detectors, radar, reported incidents, visual observation
- Not a comparison of vendors’ products

“Virtual Test Bed” Deployment Sites



Deployment Conditions

Tested in “Real World” Conditions

- Existing camera infrastructure
- Typical TMC practices and workflow



Conditions Not Controlled to Ensure Optimum Performance

- Camera settings & system configurations not always ideal for video processing (doing this could affect viewing ability)
- Normal panning/zooming of cameras
- TMC operations did not allow for constant monitoring and re-configuring of Video Analytics. Efforts made to adjust systems as much as practical.

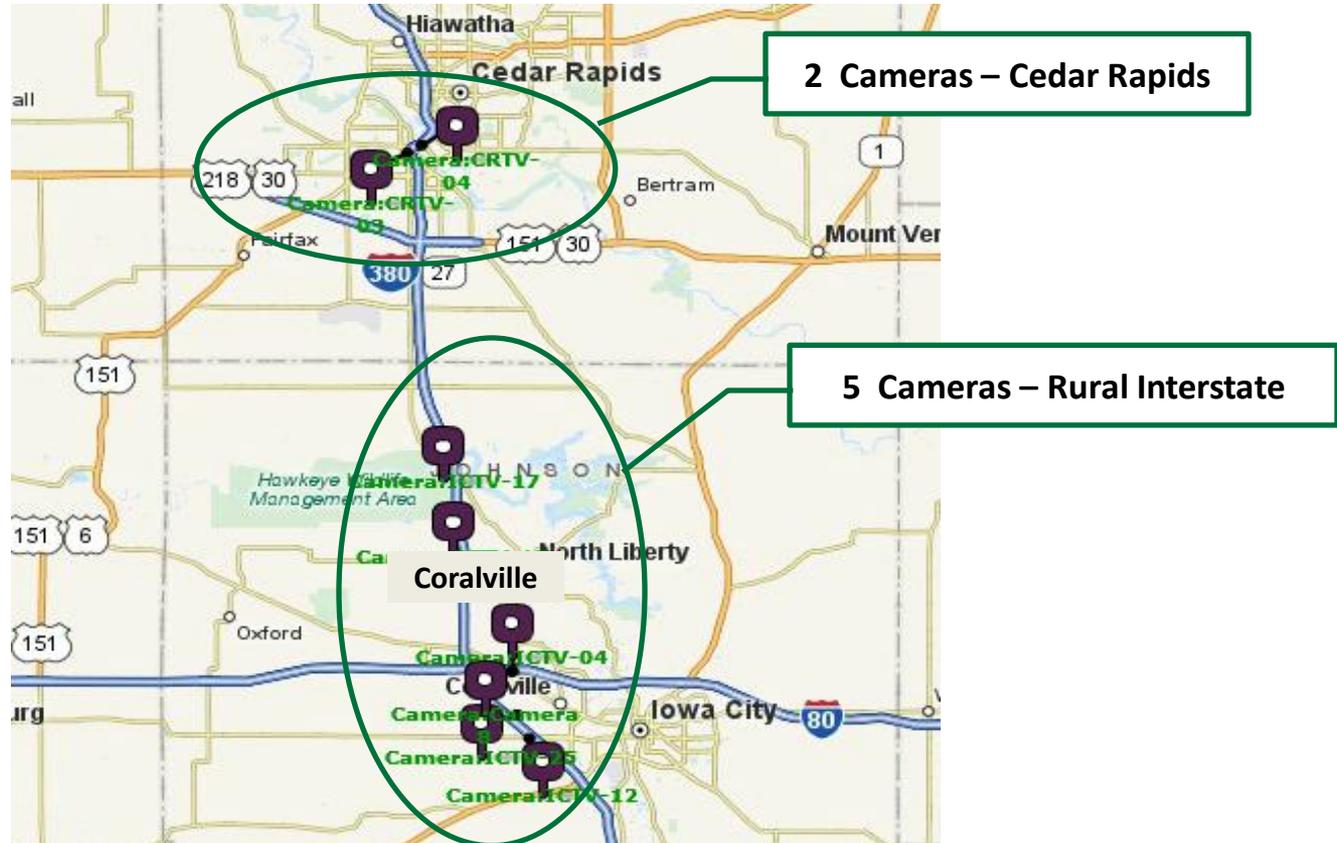
INCIDENT DETECTION



Incident Detection

Cedar Rapids - Rural Deployment

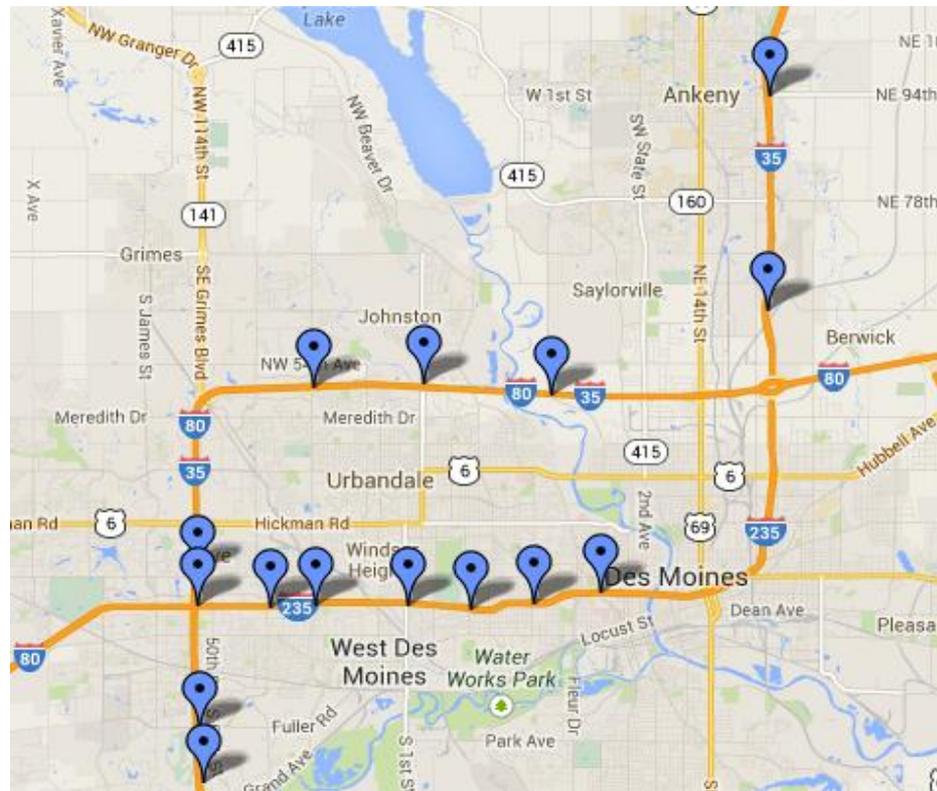
7 cameras instrumented - 2 vendors



Incident Detection

Des Moines Deployment – Urban / Suburban
15 cameras instrumented – 1 vendor

(Approx. 12% of Des Moines freeway network “coverage” with Video Analytics)



Incident Detection

Variation in Camera Views (examples)



Incident Types Detected by Video Analytics

- Stopped Vehicle / Debris in Road
- Slow Traffic / Congestion
- Pedestrian
- Wrong-Way Vehicle

Incident Detection

Analysis Approach:

- 1) Reviewed Detection Alerts: Still Images / Video Clips
- 2) Classified Alerts:
 - Likely Detection (validated)
 - Detection Not Likely (not validated)
 - Unable to Determine
- 3) Calculated % validated, % not validated , % unable to determine (as a function of total number of alerts)
- 4) Highest level of performance reported

Incident Detection

Examples - Incident detection validated

Stopped Vehicle



Incident Detection

Examples - Incident detection validated

Stopped Vehicle

North I-380 @ 1st Ave (CRTV03) 2013-11-14 21:14:40



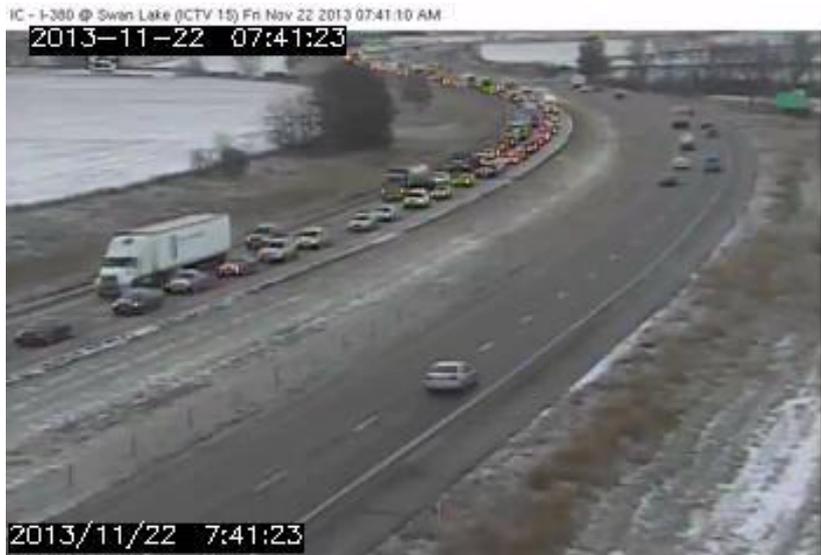
South I-380 @ 4th St (CRTV04) 2013-11-19 19:29:14



Incident Detection

Examples - Incident detection validated

Slow Traffic / Congestion



Incident Detection

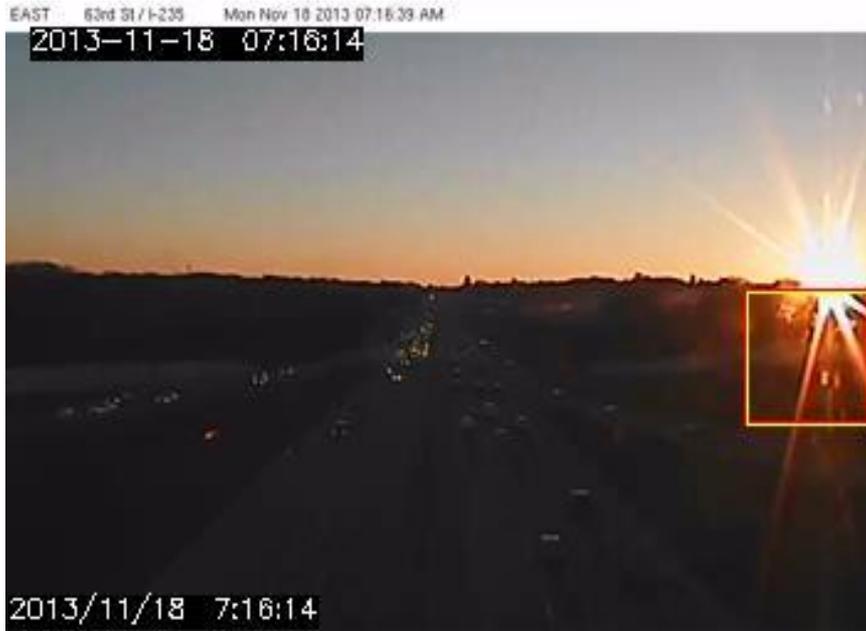
Example - Incident detection validated

Pedestrian detected as “Stopped Vehicle”



Incident Detection

Examples - Incidents not validated (false alarms)



Incident Detection

**Examples – Incidents not validated
(False Alarms caused by Obstructions in View)**



Incident Detection

Examples - Unable to determine



DMTV50 Thu Nov 21 2013 10:34:38 AM
10:34:37



Incident Detection

Results:

Highest Level of Performance

Stopped Vehicle / Debris:

72% alerts validated, 23% not validated, 5% unable to determine
(81 alerts during a 44-day period)

Stopped Vehicle / Debris – Remove False alarms from Object in View:

0% “false alarms” (26 alerts during a 21-day period)

Slow Vehicle/Congestion:

30% alerts validated, 33% not validated, 37% unable to determine
(1111 alerts during a 44-day period)

Pedestrian in Road:

None observed

Wrong-Way Vehicle Movements:

None observed

Incident Detection

Results

Factors that Impacted Performance

- Objects in the field of view
- Weather events / moisture on camera lens
- Headlight glare on roadway during nighttime lighting conditions

Factors that Did Not Appear to Impact Performance

- Camera position (zoom level, angle to roadway)
- Inaccurate configuration of Video Analytics to roadway lanes (e.g. camera panning)



Incident Detection

Comparison of Detection Alerts to Agency Reported Incidents

- It is likely that Video Analytics detected a number of incidents that were not observed by agency staff, indicating that Video Analytics can be an effective tool for supplementing existing mechanisms to alert operators
- Strategic selection of camera locations along a coverage area will optimize usefulness of Video Analytics



TRAFFIC DATA COLLECTION: Iowa/Kansas City Deployments



Traffic Data Collection

Traffic Data Types:

- Volumes (Vehicle Counts)
- Average Speeds
- Vehicle Classifications

Classification Categories from Video Analytics	Corresponding FHWA Classifications
Motorcycles	Classifications 1
Cars	Classifications 2-3
Small Trucks	Classifications 4-7
Large Trucks	Classifications 8-13

Traffic Data Collection

Analysis Approach

- Data collected in 15-minute increments
- Video analytics outputs compared to outputs from DOT detectors (loops and radar)
- Absolute Percent Difference (Abs % Diff) Calculation:
 - Calculate 15 min. period difference from DOT data
 - Convert it to absolute difference (remove any '-')
 - Compute Percent Difference
 - Result is Abs % Diff.
- Caveat: Night-time traffic is often very low volumes. Abs % Diff. is not as meaningful

Traffic Data Collection

Results : Highest Level of Performance

(All results shown are average % Diff for one week)

Traffic Volumes:

- Day: 9% Avg. % Diff. (*carries reasonable expectation of repeatability*)
- Night: 17% Avg. % Diff. (**Does not** carry reasonable expectation of repeatability)

Vehicle Speeds:

- Day: 2% Avg. % Diff (*carries reasonable expectation of repeatability*)
- Night: 6% Avg. % Diff (*carries reasonable expectation of repeatability*)

Vehicle Classifications:

- "Motorcycles" (FHWA Classification 1): Avg. % Diff of 24% at night
- "Cars" (FHWA Classifications 2-3): Avg. % Diff of 13% daytime
- "Small Trucks" (FHWA Classifications 4-7): Avg. % Diff of 44% daytime
- "Large Trucks" (FHWA Classifications 8-13): Avg. % Diff. of 23% daytime

Traffic Data Collection

Results

Factors that Impacted Performance

- Low light / dark conditions
- Camera position (proximity to traffic, zoomed out, angled to roadway)
- Weather events that reduce image quality
- Inaccurate configuration of video analytics to roadway lanes
- Camera settings (e.g. shutter speed, max gain)

Factors that Did Not Appear to Impact Performance

- Position of camera relative to direction of traffic (e.g. counting headlights vs. tail lights at night)

TRAFFIC DATA COLLECTION:
Ontario Ministry of Transportation (MTO)
Deployment



Traffic Data: MTO Deployment

MTO Deployment – Focus on Volumes

- 13 cameras instrumented at 4 Locations
- Data collected in 15-minute periods
- Video recorded for 1 week at each camera, sent to video analytics vendor for processing
- Manual counts conducted for comparison
- Manual counts compared to video analytics data outputs to compute percent error

Traffic Data: MTO Deployment

Results:

Type of Comparison	Configuration/Setting	% Error
Time of Day	Day ¹	9.1%
	Night	7.9%
Camera Angle	Side	9.4%
	Overhead	6.5%
Camera Type	Axis	7.5%
	Cohu	9.6%

¹ 'Day' analysis was PM peak (16:30-17:30)

Traffic Data: MTO Deployment

Results:

1. Camera based counting system is appropriate if:
 - Overall Accuracy within 10% is acceptable
 - Vehicle Classification is not critical
2. Camera based counting system may not be suitable if:
 - Counts are to be conducted in work zones or areas with high stop-and-go traffic
 - Accuracy within 5% is required
 - Vehicle Classification is needed
 - Night-time accuracy is important

Traffic Data: MTO Deployment

Lessons Learned:

1. Engage in discussions early with camera vendors
2. Standard definition cameras are actually better
3. Ambient light surrounding cameras should be taken into consideration for camera locations

Next Steps:

MTO will be undertaking additional data collection assignments utilizing video analytics beginning this fall and continuing through next summer

WRONG-WAY VEHICLE DETECTION



Wrong-Way Vehicle Detection

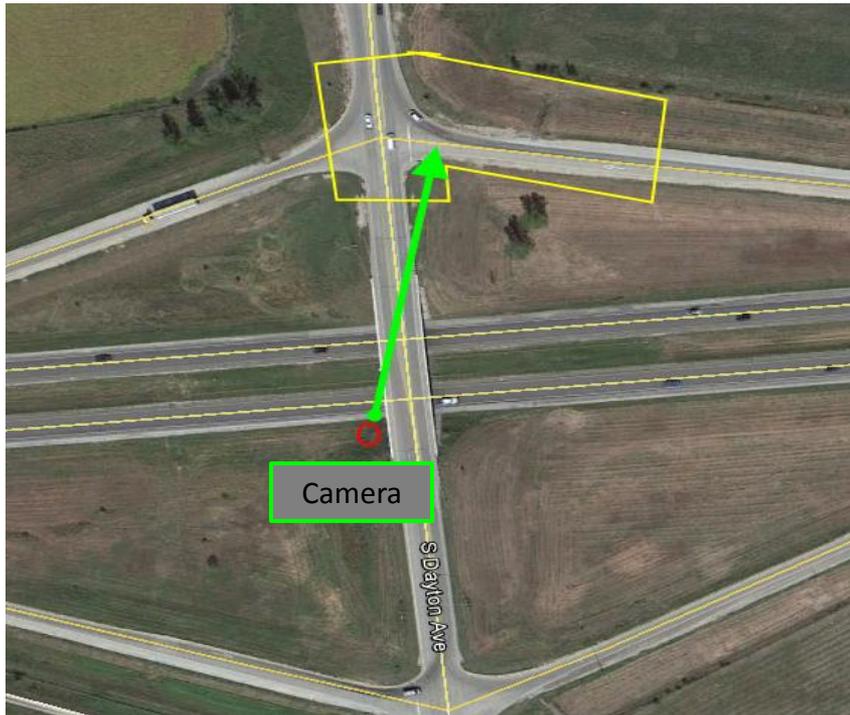
Controlled Test: Nov. 2013 in Ames, IA

- 3 vendors/technologies at 3 separate freeway ramps
- Ramp closures to test various conditions
- Detections conveyed via email, web interface, or on-site computer interface
- Recorded “detection” or “non-detection”



Wrong-Way Vehicle Detection

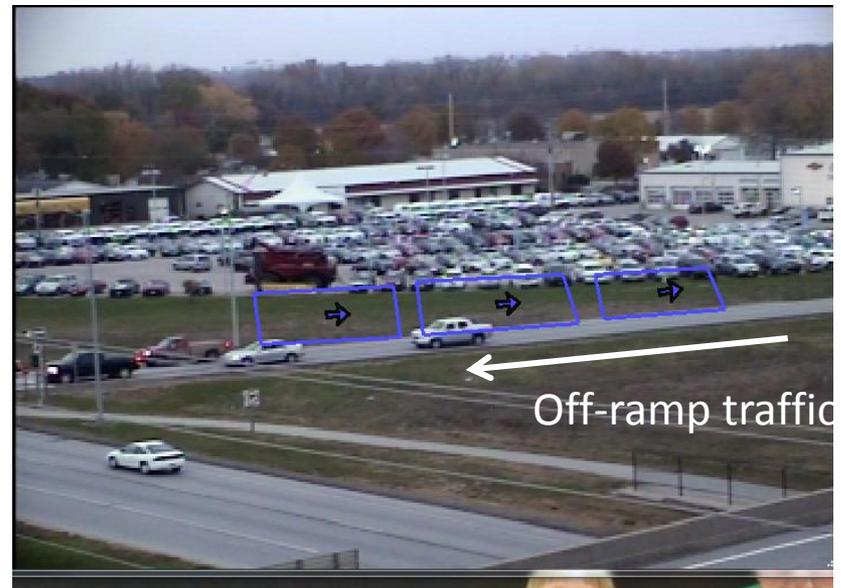
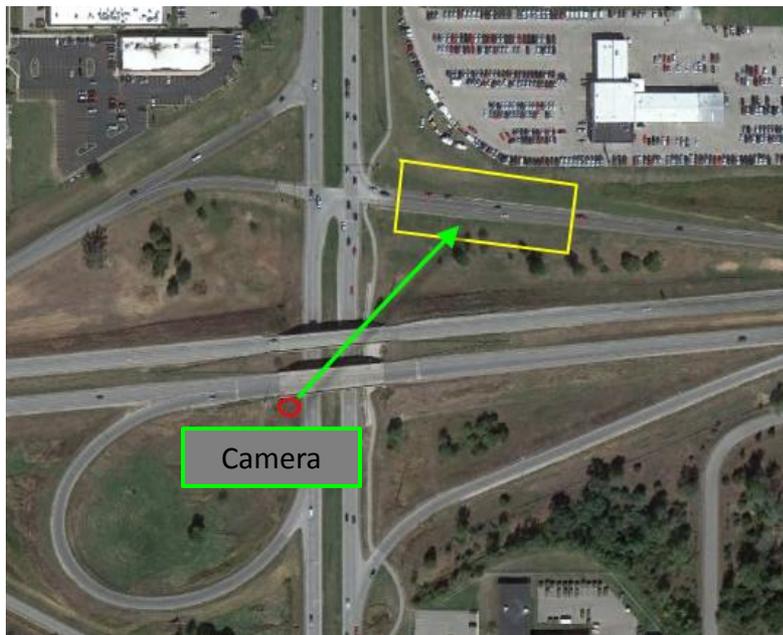
Deployment Site #1 US 30 at Dayton Ave.



90 degree detection

Wrong-Way Vehicle Detection

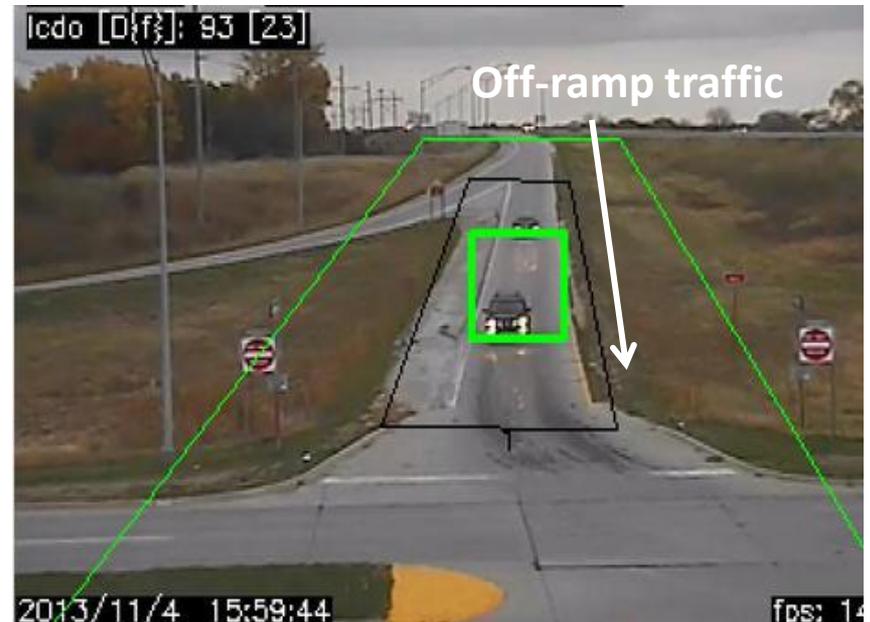
**Deployment Site #2
US 30 at Duff Ave.**



90 degree detection

Wrong-Way Vehicle Detection

Deployment Site #3 US 30 at University Blvd.



"head-on" detection

Wrong-Way Vehicle Detection

Highest Level of Performance Achieved
Daytime Test: 100% detection for 12 test drives Nighttime Test: 83% detection for 12 test drives
Factors that Impacted Detection Rate
Nighttime / Low Light Conditions Slow Speeds
Factors that Did Not Appear to Impact Detection Rate
Color/Size of Vehicle Lane Position (consistent position, shoulder, and/or weaving)

LESSONS LEARNED



Lessons Learned

Planning and Procurement

1. Determine Uses and Needs

- What will the system be used for?
- What are the most important uses (e.g. traffic data collection, incident detection, etc.)?

2. Understand Limitations of Multi-Purpose Capabilities

- Camera positions / settings may serve one application better than others
- Multiple uses may be difficult or impractical

Lessons Learned

Planning and Procurement, cont'd

3. Recognize Investment Tradeoffs

- Potentially lower up-front investment with Video Analytics
- Consider continuing costs: Staff training, setup, and ongoing monitoring/configuration

4. Utilize Fixed Cameras and/or Dedicated Cameras for Traffic Data

- Traffic data tends to be more accurate with cameras that remain stationary (fixed, dedicated)
- Consider installing temporary dedicated cameras where infrastructure does not allow optimized positioning

Lessons Learned

Planning and Procurement, cont'd

5. Optimize Video Feed Quality and Communications

- Video feeds with minimal interruption are desired. “Choppy” feeds/communications will not be accurately processed.
- Ask vendors to provide feedback on feed quality
- Test video feeds in advance of procurement

6. Include Design & Testing Provisions in Procurement

- Add tasks for additional testing and tuning 6 months to 1 year after initial deployment

Lessons Learned

Planning and Procurement, cont'd

7. Make 'Go/No-Go' Decisions When Selecting Cameras

- Work with vendors to determine if camera positions are suitable

8. Consider Future Potential for Video Analytics when Installing New Cameras

- Even if Video Analytics deployments are not planned, consider potential for future use when installing new camera infrastructure

Lessons Learned

Deployment

1. Dedicate Agency Resources to Deployment Activities

- Agency resources needed during installation and troubleshooting during set-up
- Schedule check-in visits with vendors

2. Commit to Learning & Understanding System Procedures

- Dedicate resources to learning system configurations, procedures, and performance impacts
- Operators should fully understand capabilities to ensure that the system is as useful and accurate as possible

Lessons Learned

System Operation

1. Use Camera Presets and Auto-Return to Preset Positions

- Cameras should reset to optimal Video Analytics positions after being manually moved

2. Monitor Calibrations and Adjust as Needed

- Operators should ensure that cameras are returned to their optimal view settings (use presets, if possible)

3. Recognize Strong Link Between Human Interaction & System Performance

- Success is dependent on agency's level of commitment
- Resources needed to monitor performance, adjust and re-configure when cameras pan/zoom, etc.

Lessons Learned

Evaluation

1. Establish Performance Parameters

- Develop subjective “success” parameters to determine if a system performs to pre-determined standards

2. Compare/Contrast Video Analytics to Other Detection Mechanisms

- Compare performance outcomes of various technologies for specific uses

3. Extend Incident Detection Testing to “Missed Incidents”

- Determine extent to which Video Analytics fails to detect actual incidents
- Utilize closed test track or other controlled environment

EVALUATION FINDINGS



Evaluation Findings

State of Practice for Video Analytics is ready to meet many agency needs.

- Dedicated and/or fixed cameras may be warranted, especially for traffic data collection
- Video Analytics may not serve all purposes simultaneously (e.g. a camera used for incident detection may not be optimal for traffic data collection)
- Important to follow vendor guidelines for camera selection, position, zoom level, etc.
- Recognize significant human component involved. Operator resources are required to monitor system settings and re-configure as needed.

Acknowledgements & Project Contact

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