

Novel uses of smartphones in transportation

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11/01/2015

Smartphone Revolution

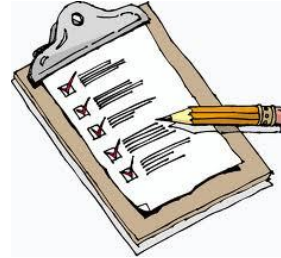


Nice for entertainment, especially for the kids...



Smartphones for infrastructure monitoring

Road damage



Lane marking retroreflectivity



Traffic sign retroreflectivity



Consumer Sensors



Smartphone

\$300

Camera

GPS

Wi-Fi

Accelerometer

Compass

Computing

3G/4G

Vehicle State

OBDII

\$15

Bluetooth or Wi-Fi

Ignition on/off

Speed

Temperature

Barometric pressure

RPM

Many more



Samsung Galaxy Camera

\$500

Android (everything but the phone)

16 MP => 4 Hz for 20 images

1040p => 30Hz

640x480 => 120 Hz

21x zoom

Optical stabilization

Select ISO, exposure, aperture

Gyro





The inventory of public infrastructure maintained by the streets/parks Maintenance Division includes:

- **866 lane miles of asphalt streets.**
- **90 lane miles of concrete streets.**
- **80 lane miles of brick and block stone streets.**
- **620 signalized intersections**
- **655 sets of city steps, covering 22 lineal miles.**
- **4400 street lighting fixtures.**
- **850,00 street signs.**
- **1,672 City lots part of parks, greenways and City government facilities.**
- **7,600 City, County, and Board of Education jointly owned lots.**
- **195 park facilities and various green spaces.**
- **330 Athletic Courts**
- **128 Athletic Fields**
- **134 play areas consisting of playgrounds, parklets, and tot lots.**

Goal of proposed system



Monitor road surface damage

- inexpensively**
- accurately**
- continuously**

Mounted in car



Smartphone mounted
on the windshield

Powered from
cigarette lighter

Install it on any vehicle that regularly drives in the city



Low Cost

- Consumer Device



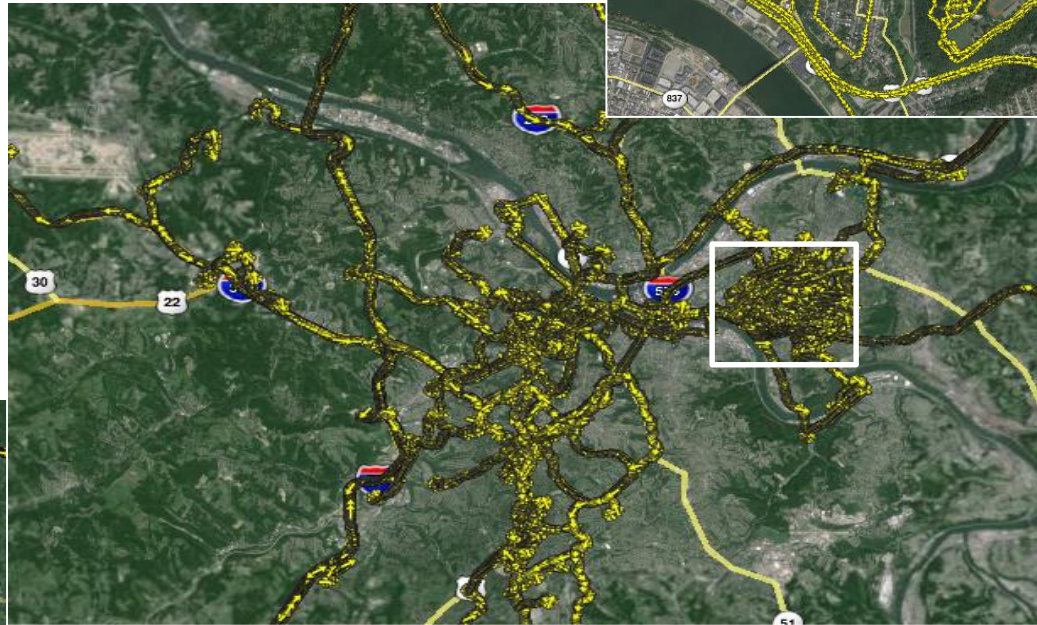
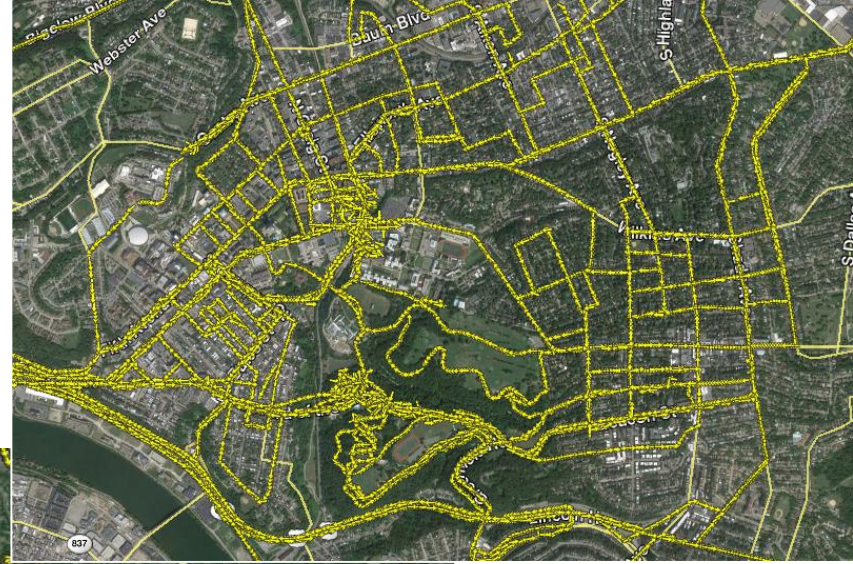
- Free/inexpensive/existing software and services
 - Dropbox, Google Earth, Open Source software, GIS
- Minimal labor
 - Collection by vehicle that drive for other purposes
 - Automate processes and analysis

Example Input Video - 10 Hz



Data overview

2 years
250 hours
11k videos
7M images
1000 km unique roads



CMU



Pittsburgh

Pennsylvania

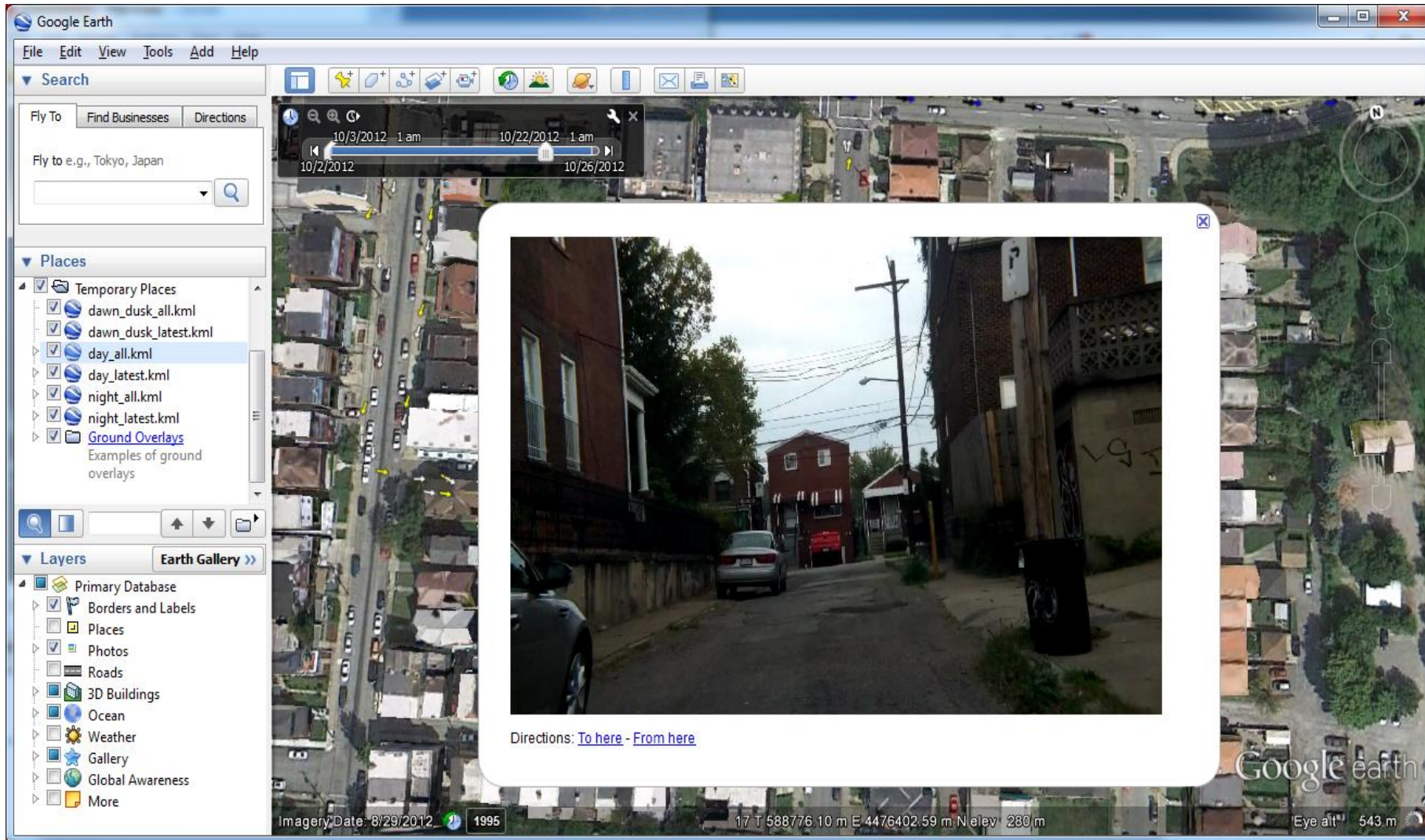
Display: Google Earth

Select time window



- Day latest
- Day all
- Night latest
- Night all
- Dawn/dusk latest
- Dawn/dusk all

Display raw images



Automatic Road Damage Detection





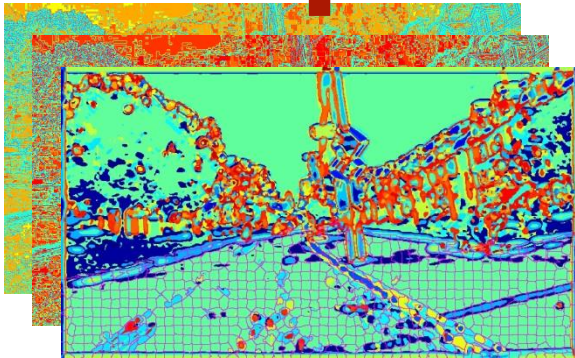
step 1:
identify ground pixels



step 2:
compute SLIC
superpixels



step3:
compute features on
superpixels₁₆



step 4:
train/run a SVM
classifier on
superpixel
features





Final Result



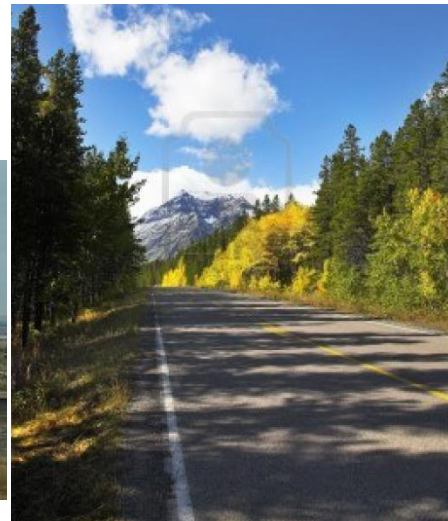
no damage

some damage

lots of damage

Computer Vision is difficult!

- Rule of thumb:
 - 80% detection is doable
 - 95% detection is very hard
 - ~~99% is almost impossible~~
 - 99% needs deep learning and lots of labeled examples (100K)



Advantage of lots of data

- Select data during the day when it was overcast (good illumination, no rain, no shadow)
- Select data during spring and summer (no snow, no leaves)
- Score the same road many times

Some human interaction needed

- Check periodically to ensure quality.
- System should flag borderline cases.
- System should flag things that don't make sense
 - Sudden changes from one collection to another
 - Things that have not been seen before

On the other hand...

- In some things computers are much better:
 - Human: judging crack vs. good road vs. shadow
 - Computer: Measuring areas

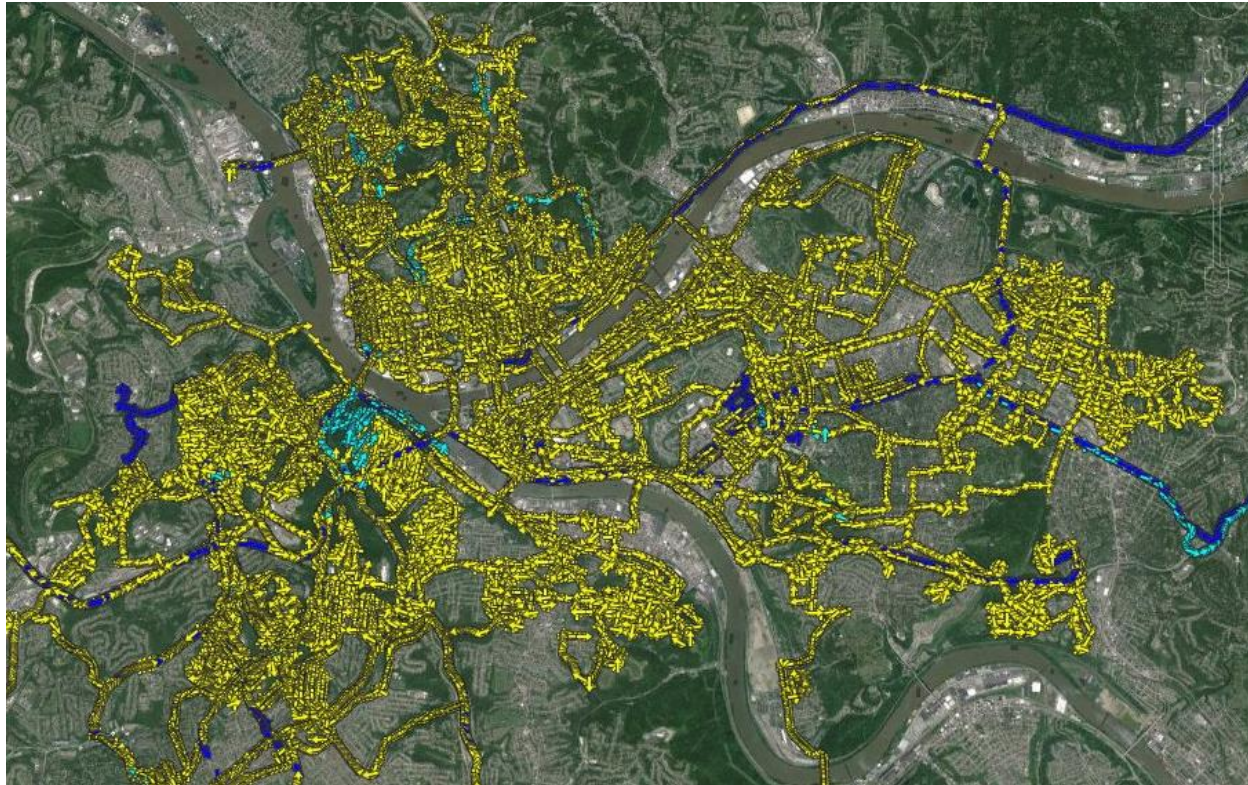


General lesson: Don't just trust the computer, work with it!



Pilot Test in Pittsburgh

- Mounted data collection system on 3 Pittsburgh vehicle
- Implement road assessment into their infrastructure management system.



Road Condition Reporting

Clear



Significant
snow cover



Wet with
freezing
conditions



Icy



Snow and/or
slush covered






impassable



Camera Mount



-  Suction Cup
-  Quick release
-  Black Felt Shield

Data Collected

- Almost 40 hours of data
- 23 shifts (some are a few minutes, others several hours)
- Two snow plows in two districts (D-4 and D-9)



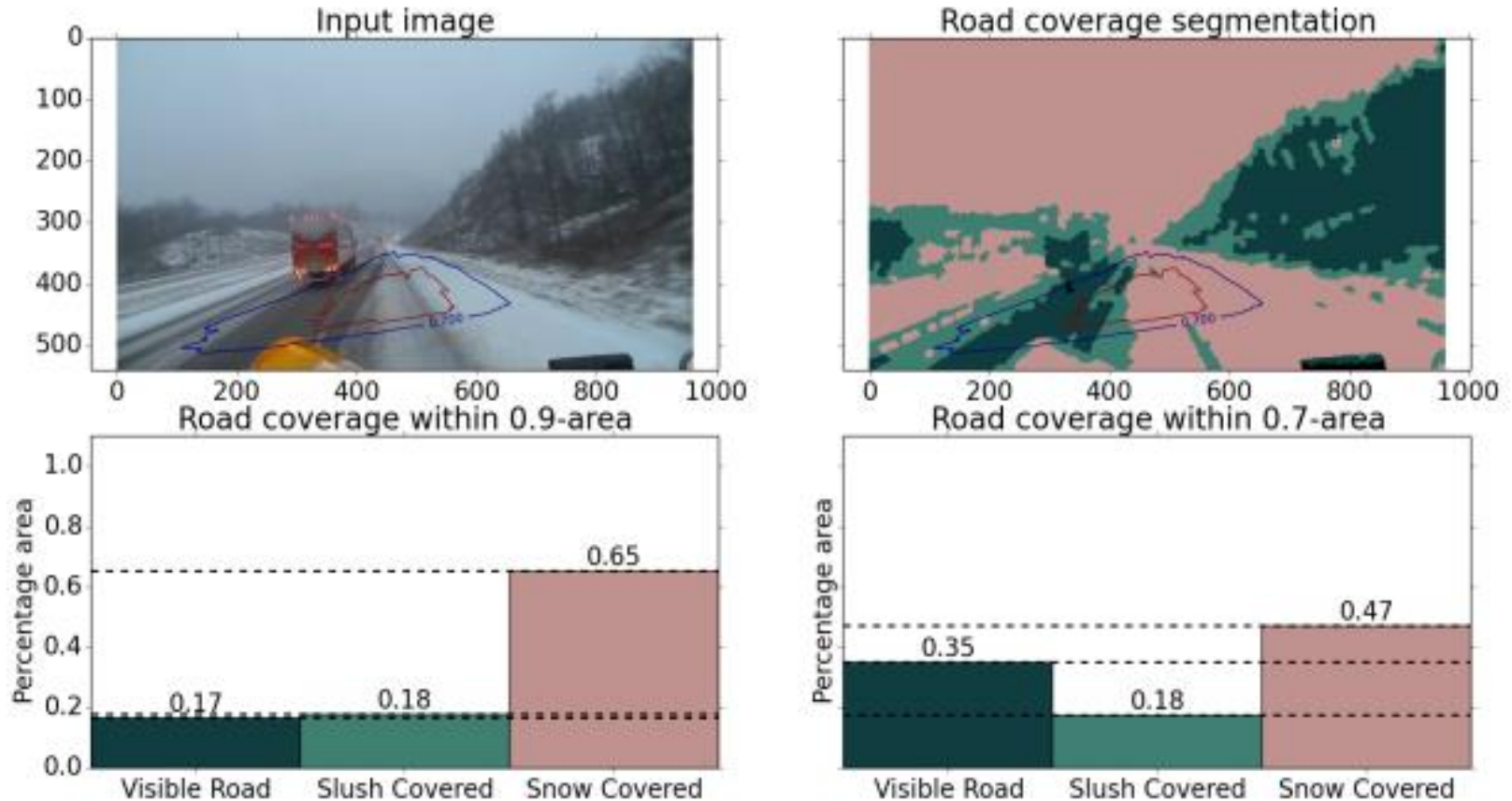
Sample Images -day



Sample Images - night



Determining road/slush/snow using computer vision



— Road directly in front of snow plow

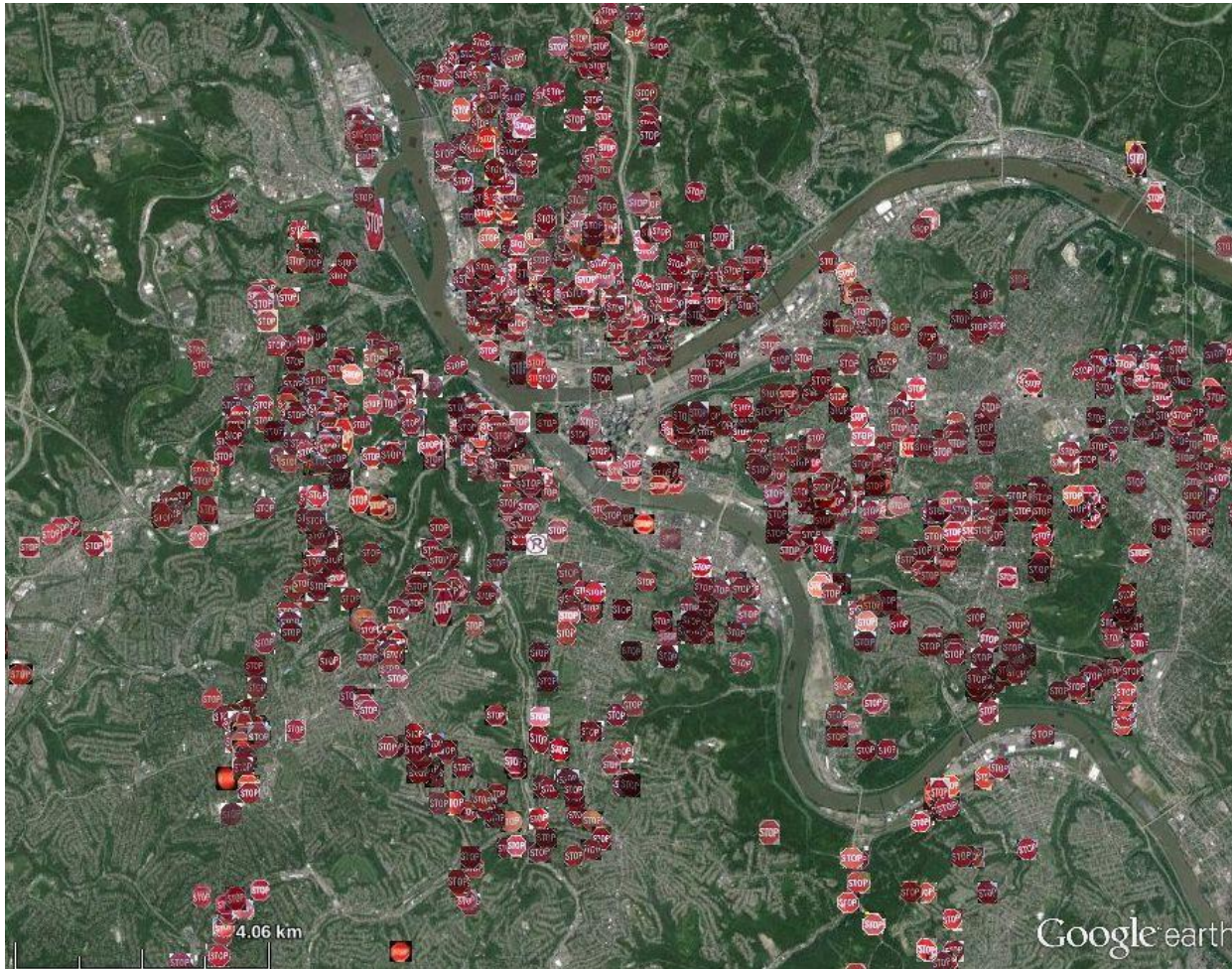
— Road including adjacent lanes

Sign detection



Stop Sign Detection

- Inventory of stop signs
- Detect problems



sticker



occluded



graffiti



displaced

Lane Marker

- Find lane markers
- Determine quality



 good
 damaged

Inspection Possibilities

Algorithm developed

Pilot Phase

Vegetation overgrowth

Wires: Tree overgrowth

weather

Signs: Damage, fading, graffiti



Building: Graffiti, code violation

Sidewalk: Damage, accessibility

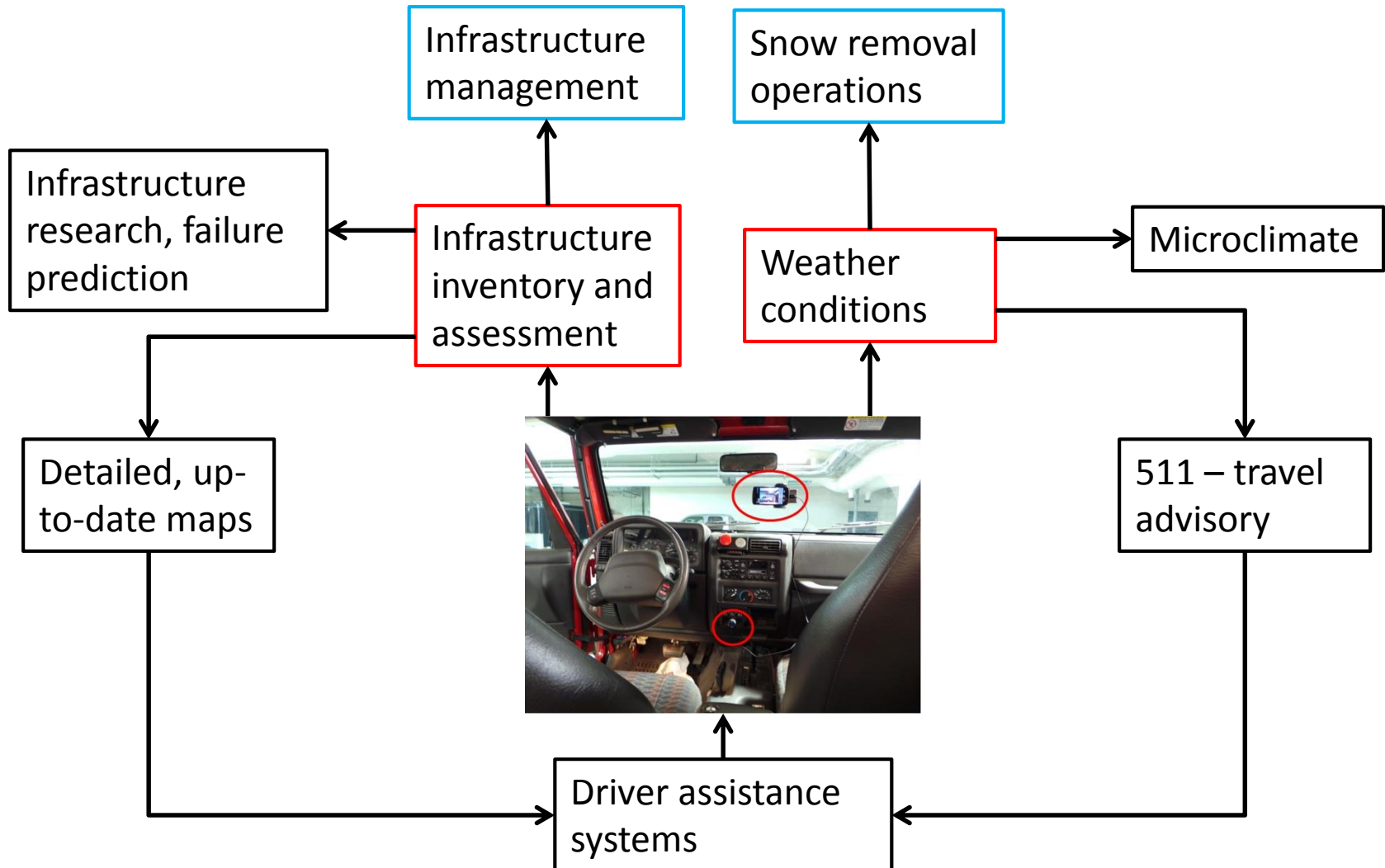
Lane markers: Damage, fading

Development started

Pavement: Damage

Drainage

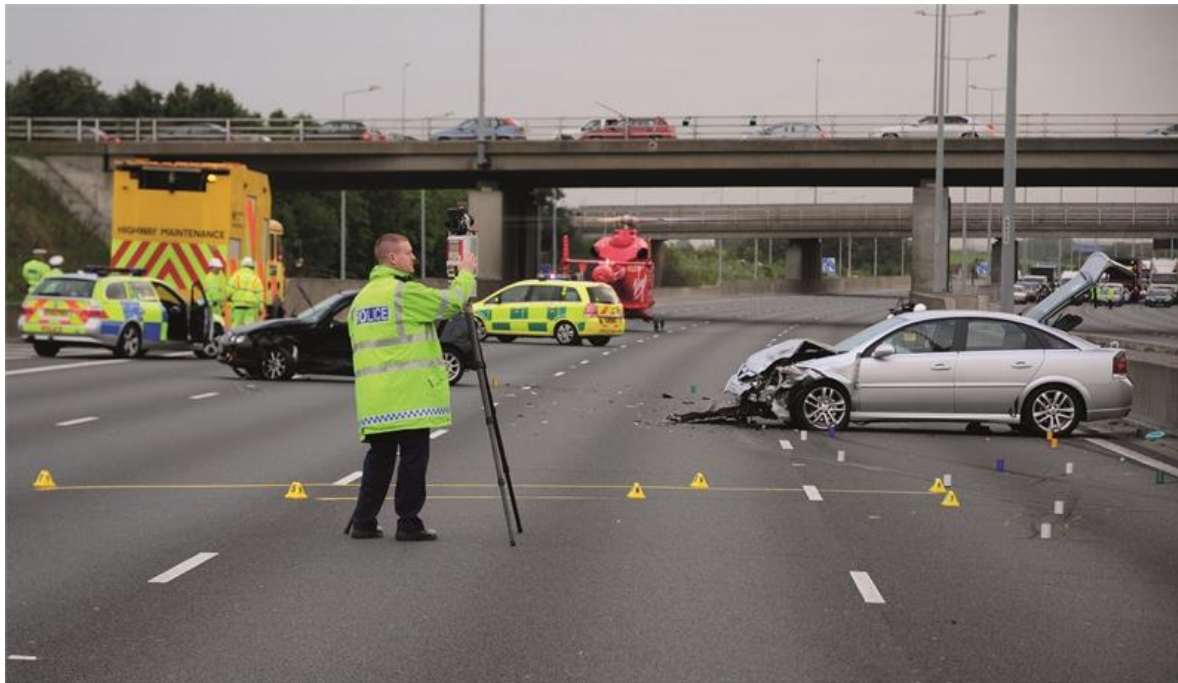
Long Term Outlook: Part of I/O devices in a car



3D reconstruction from images

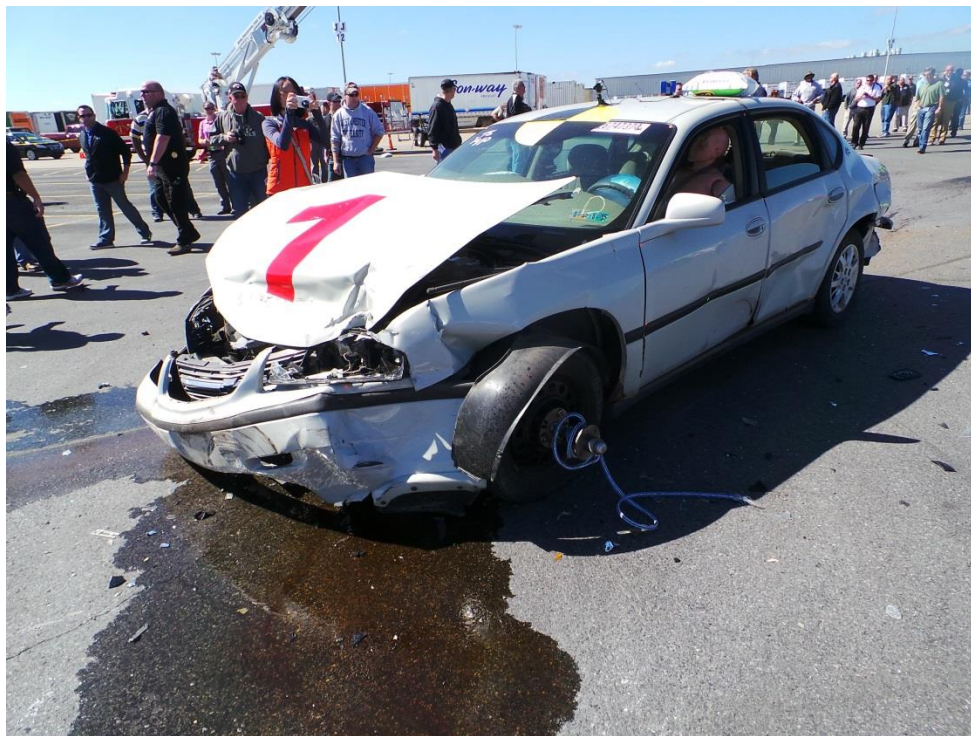
3D accident reconstruction from images

- There is very little time to record an accident scene.
- Laser scanners are very expensive.
- Currently 3D reconstruction is only done for fatal accidents:



Staged Crash





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JPG



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JPG



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JPG



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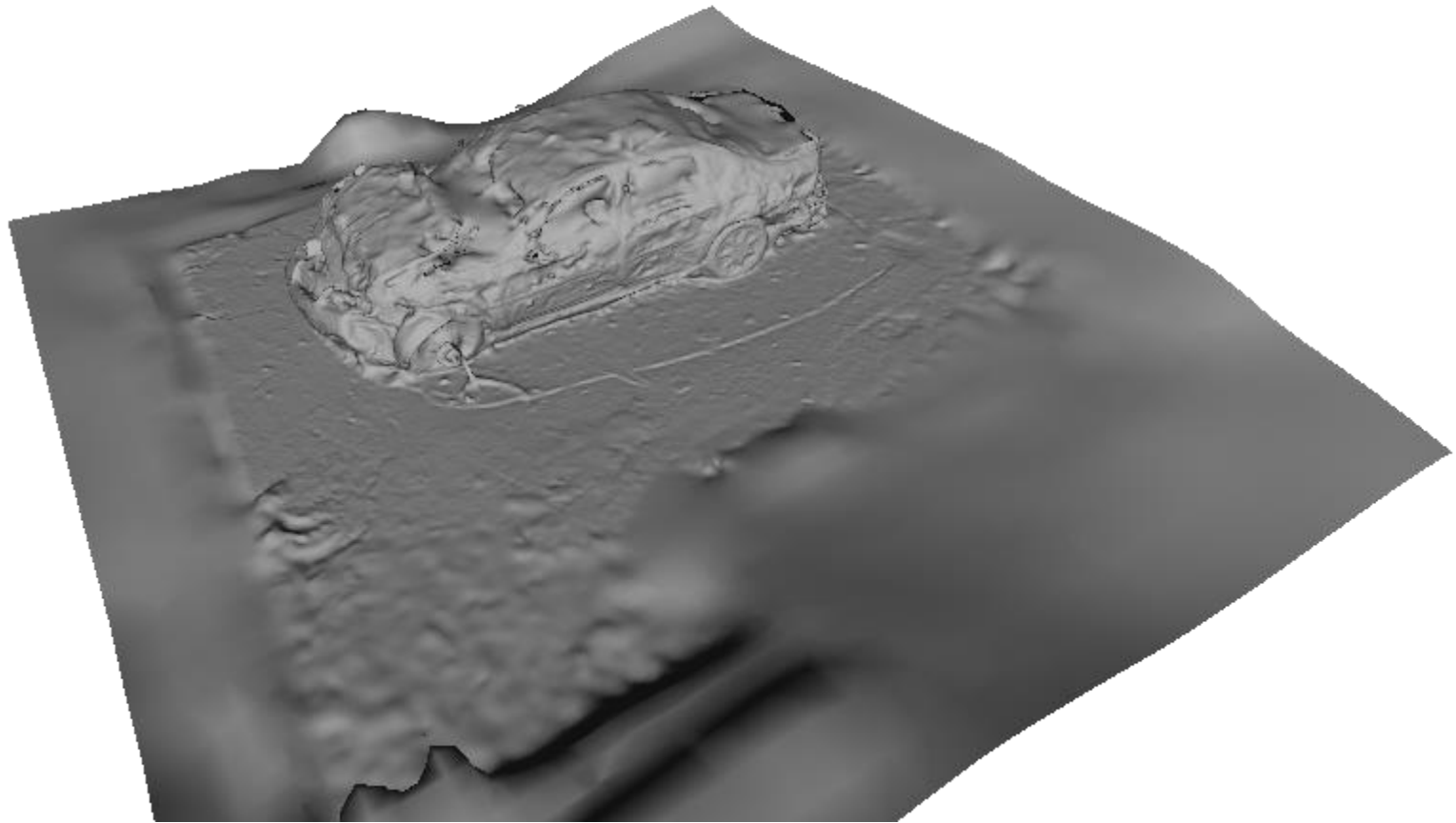


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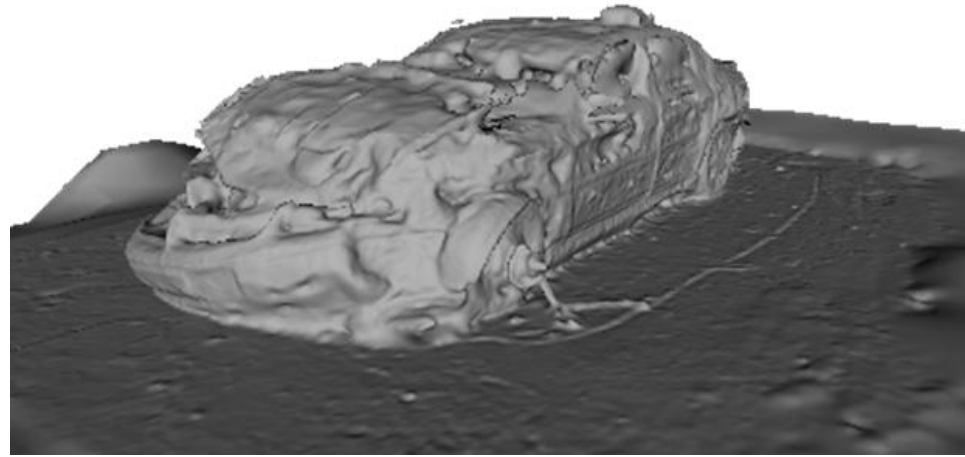
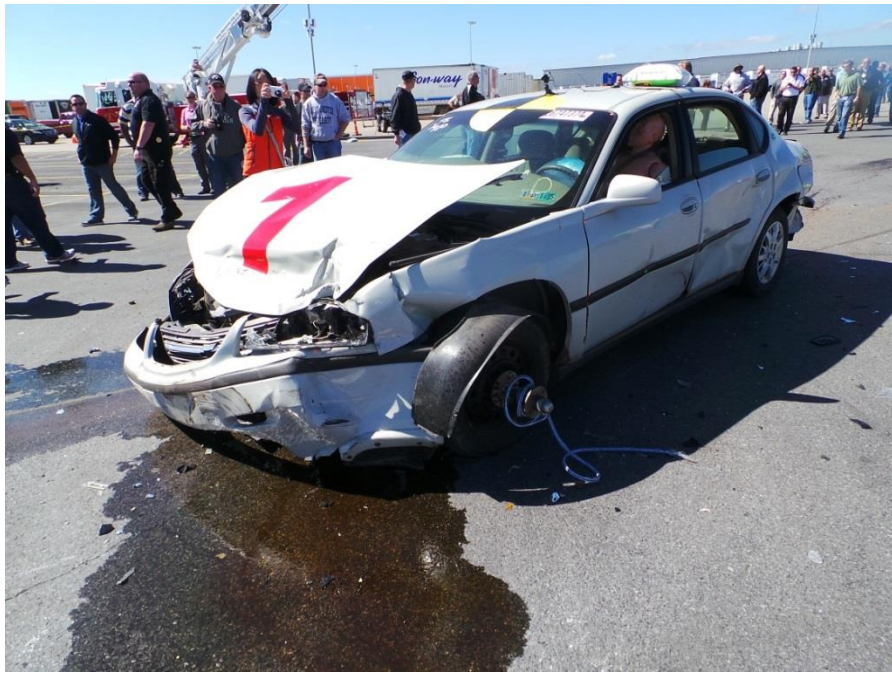
3D Model



Colorized 3D model







Before and after crash comparison



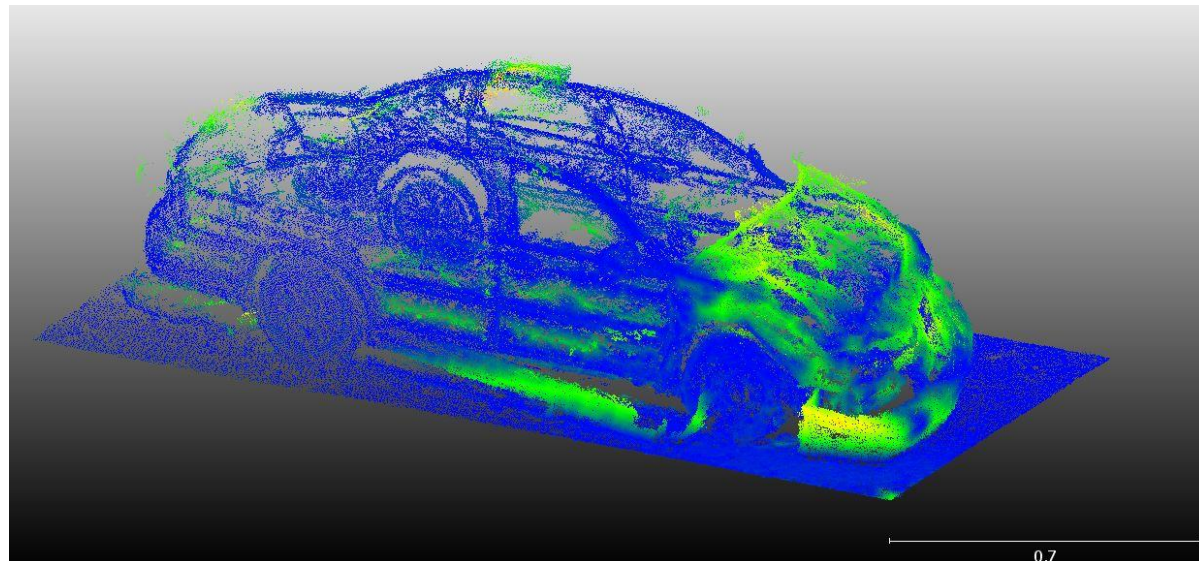
Photo before crash



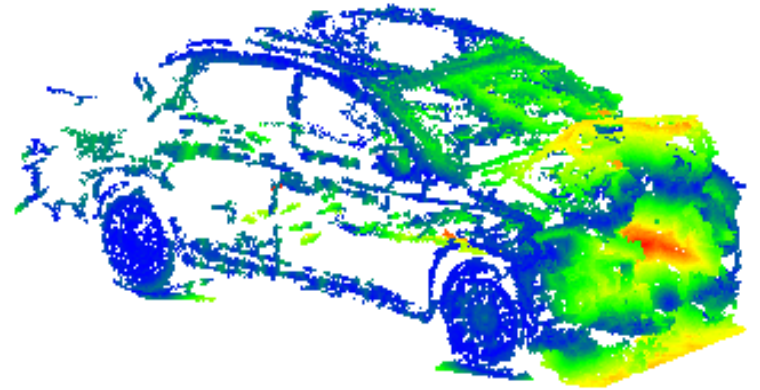
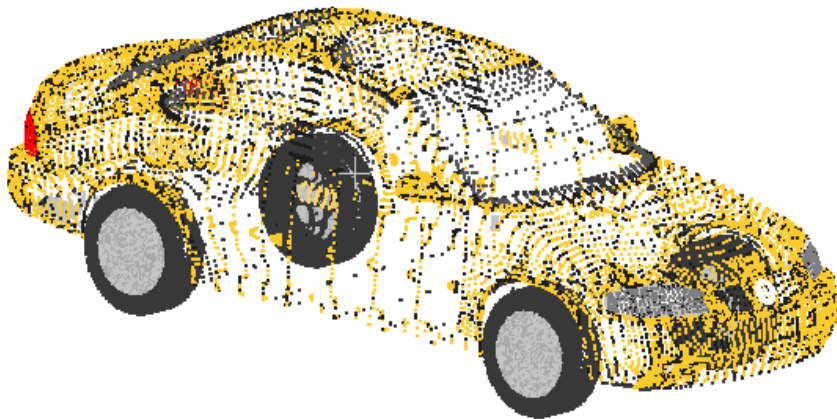
Photo after crash

The 3D models of before and after the crash are aligned and compared. The differences in the models are highlighted.

- ⇒ **Accident investigation**
- ⇒ **Crashworthiness research**

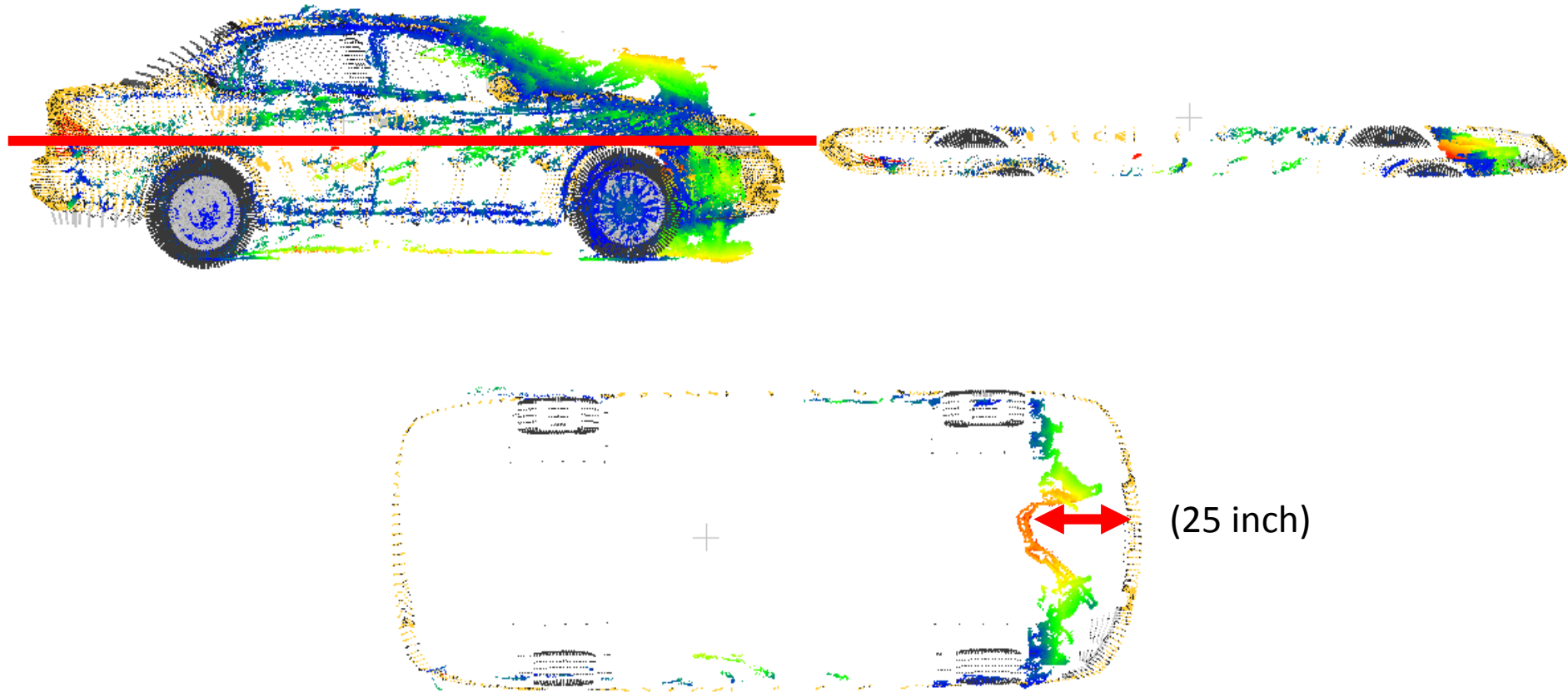


Compare crashed vs. new



Color indicates the distance of the point from the crashed car to the new car

Top-down view after aligning two models



Solid model: motorcycle crash



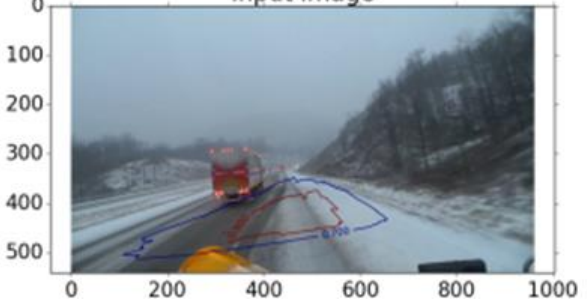
Inside & outside of a crushed car



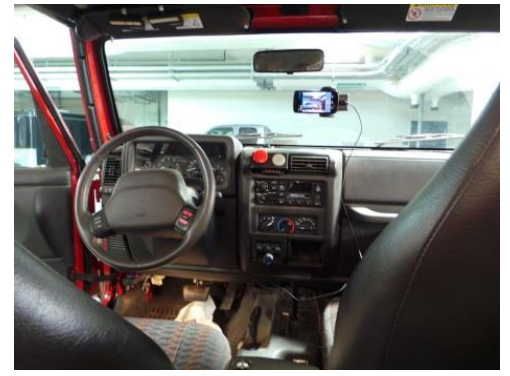
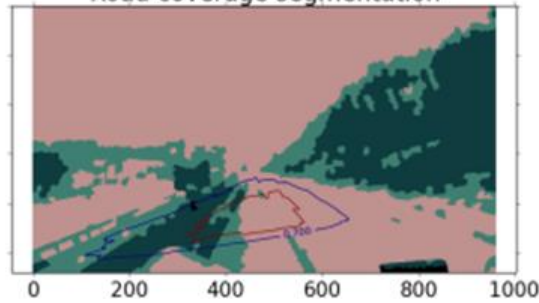
Many other possible application besides accident reconstruction



Input image



Road coverage segmentation



Thank you!

