

Mapping Land Cover and Vegetation Structure

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We need detailed information of land cover / Vegetation structure for a wide variety of applications

- There is no global or Canadian standard classification system with agreed classes and definitions
- Many municipalities develop their own classification system based on:
 - Need – How will it be used
 - Cost – How much money do they have
 - Update – Is it a one off or does it need regular updating ?



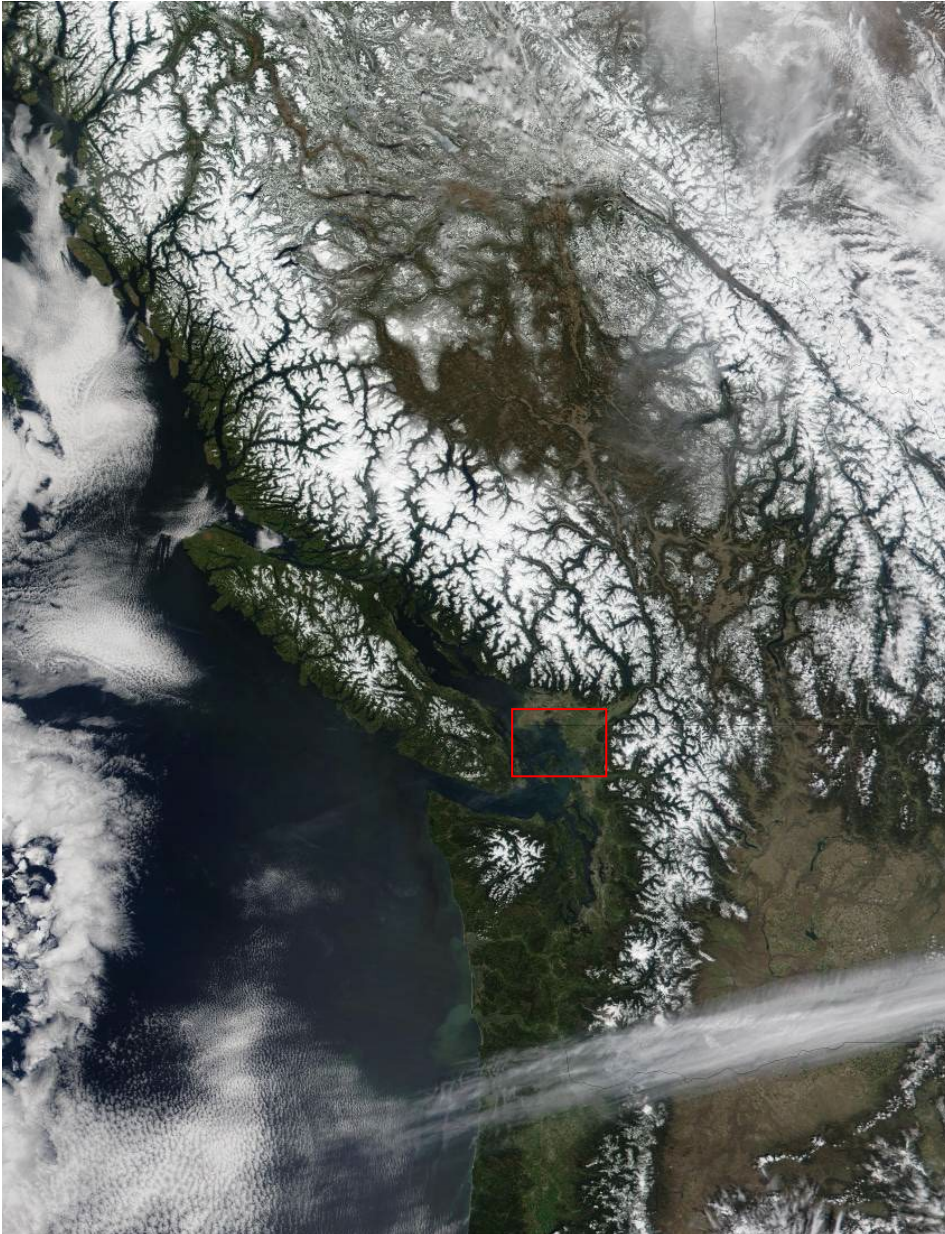
- Conventionally municipalities flew aerial imagery and manually delineated land cover types using local knowledge
 - Expensive
 - Required trained staff
 - Range of applications of the data often exceeded the data quality or design
 - Not amendable to digital analysis in GIS



- Move towards satellite data for urban and regional land cover assessment
- More recently LIDAR data provides additional information on vegetation height and complexity
- Still some classes that are not possible to map



Low spatial resolution (>100 m):



Platform: MODIS

Sensor: Terra; Aqua

Spatial resolution:

250 m – 1 km

Applications:

- Broad global land-cover types
- snow cover
- canopy cover
- sea surface temp
- vegetation phenology

Acquisition cost: free

Medium spatial resolution (<100m):



Platform: Landsat / Sentinel 2

Spatial resolution: 10 - 30 m

Applications:

- forest cover
- insect infestation
- crop forecasting
- coastal wetland erosion

Acquisition cost: free

-typically government

IA



Vancouver, Sentinel -2

High spatial resolution (<5m):



Platform:

Rapideye (no longer operations)
PLANET,
Spot 6

Spatial resolution: < 5m

Applications:

- urban mapping
- road mapping

Acquisition cost: \$

-typically private

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SPOT 6

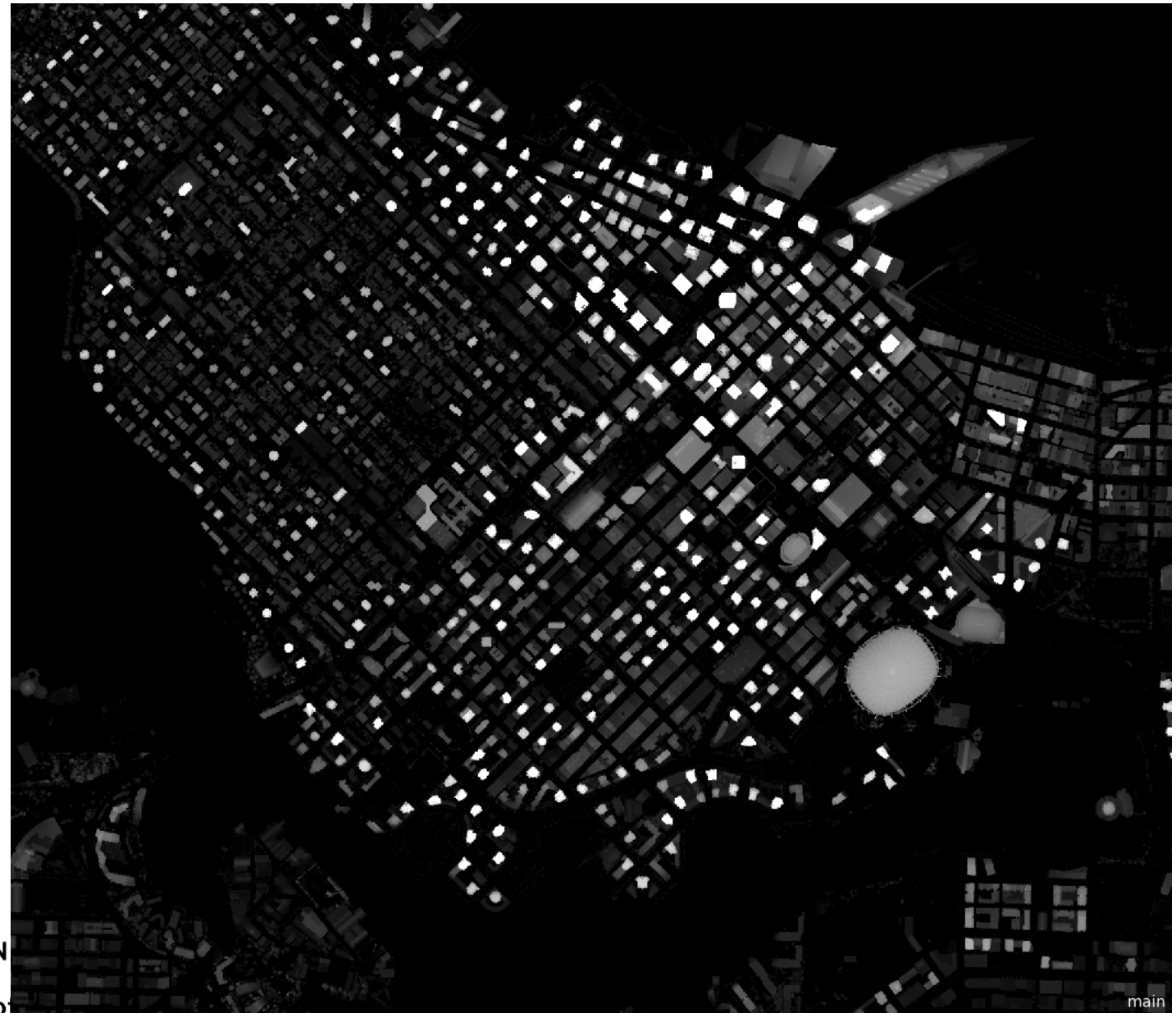




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LiDAR provides 3D structure data at high resolutions ($<1\text{m}$)

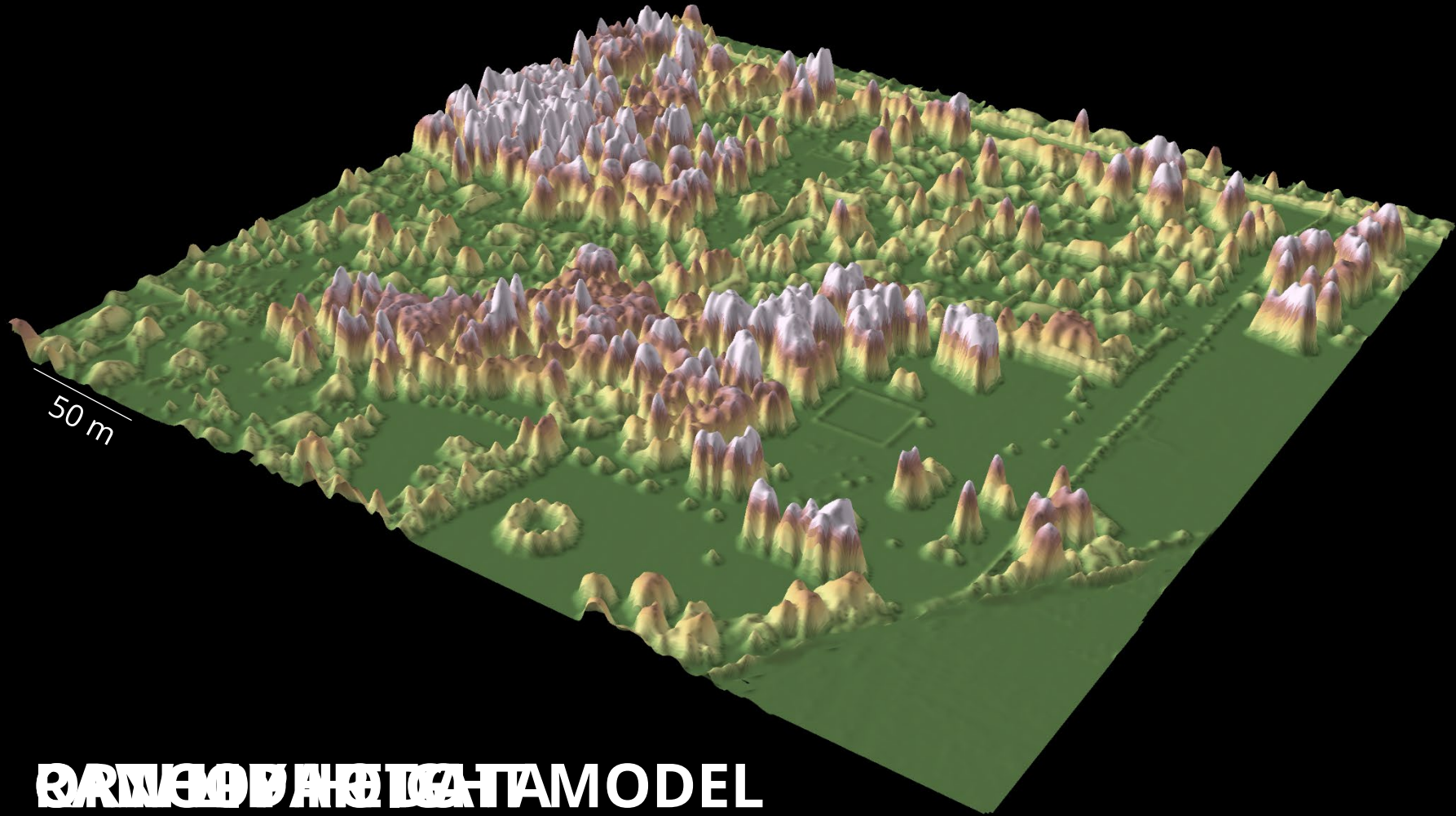
- **n**ormalized **D**igital **S**urface **M**odel (nDSM)
- Digital Elevation Model (DEM)
- Point Cloud Metrics



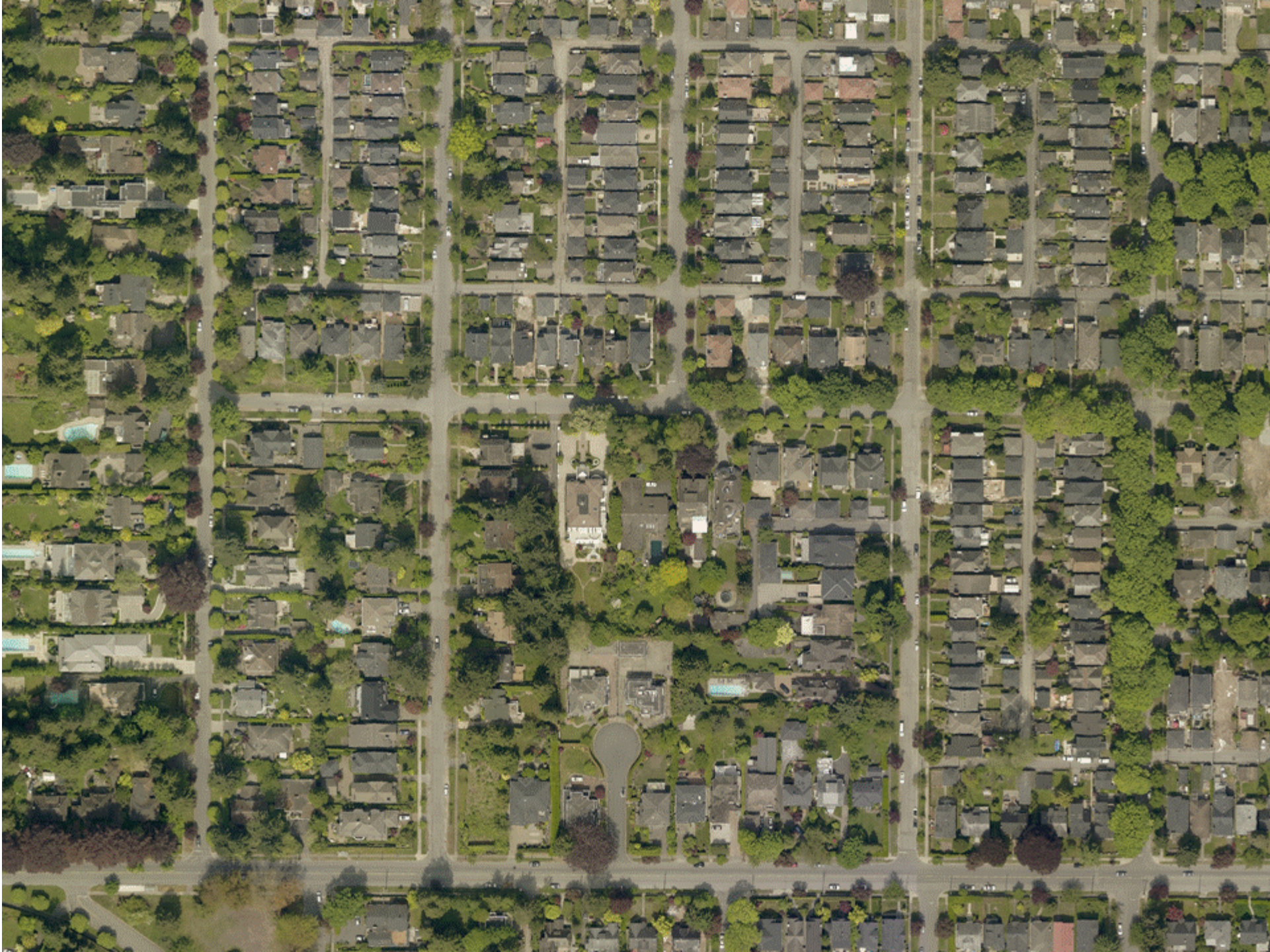








BRUNNENHOF DATAMODEL

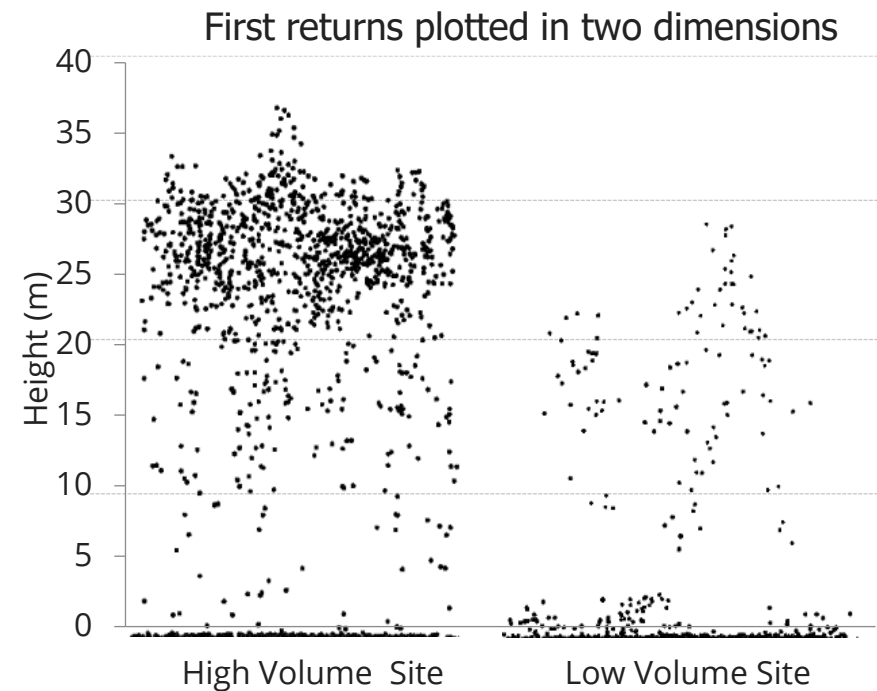
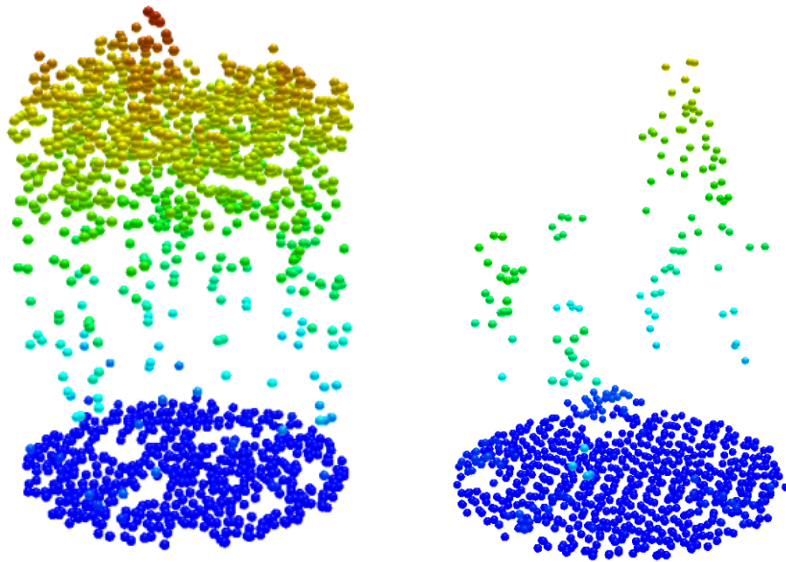


- Accurate for tall trees; misses many lower trees
- Difficult to predict species
- In reality we map clumps of trees in urban areas;
- And then in forested areas we do an Area Based Approach, typical to what we do in forestry



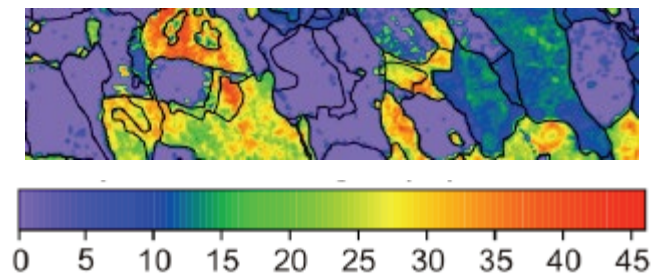
Plot Level Analysis

- Derive LIDAR metrics in a given grid size (say 10 x 10m)

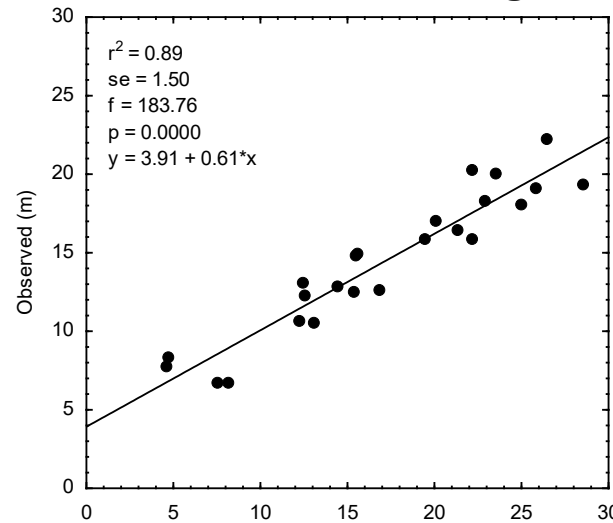


How do we relate LIDAR to ground data?

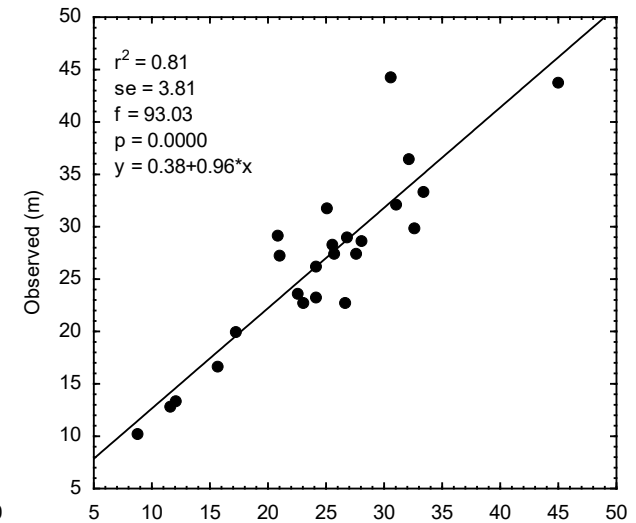
- GPS ground plot location
- Make ground measures
- Statistically relate ground measures to lidar metrics
- Can apply these relationships across all lidar grid cells (10 x 10m)
- Models can be of height, volume, biomass etc.



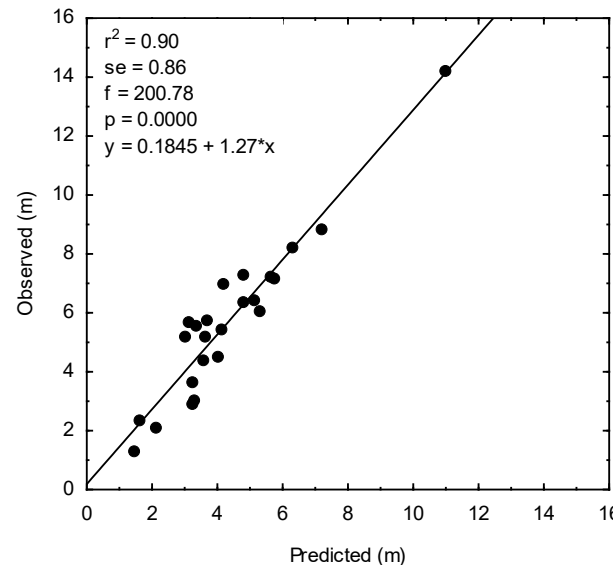
Mean tree height



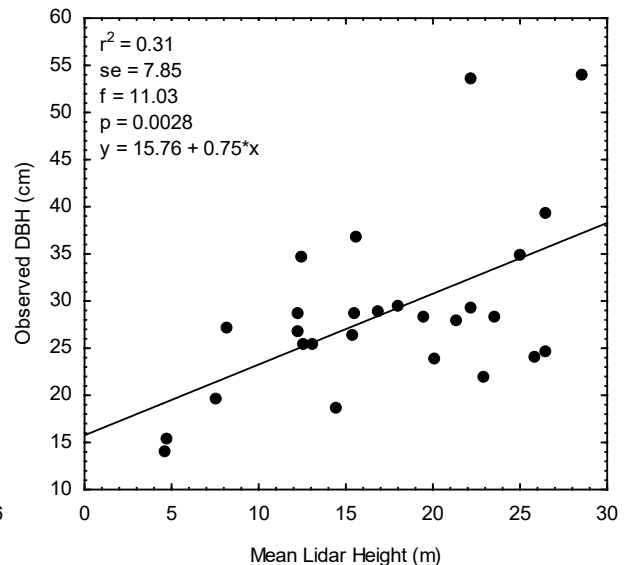
Maximum tree height



Variation of tree height



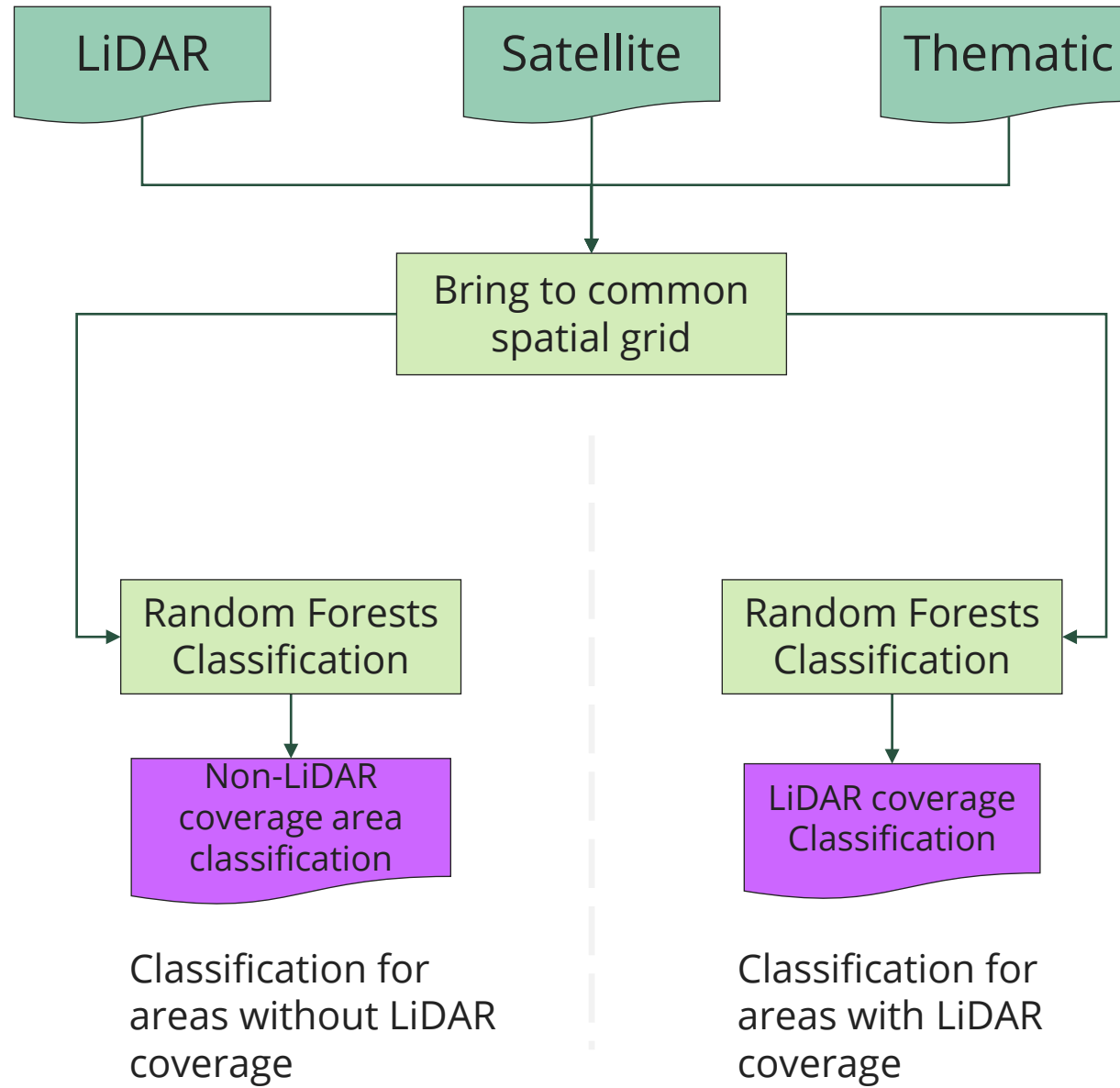
Mean diameter (dbh)



But we don't have LIDAR everywhere, it is of different dates,
densities and flown by different providers



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Built-up	Buildings	
	Paved	Linear paved
		Linear paved elevated
		Non-linear paved
Bare	Other Bare	
	Soil	
Vegetation	Tree	
	Grass	
	Shrub	
	No vegetation	
Water	Water	
Shadow	Shadow	
Clouds/Ice	Clouds/Ice	

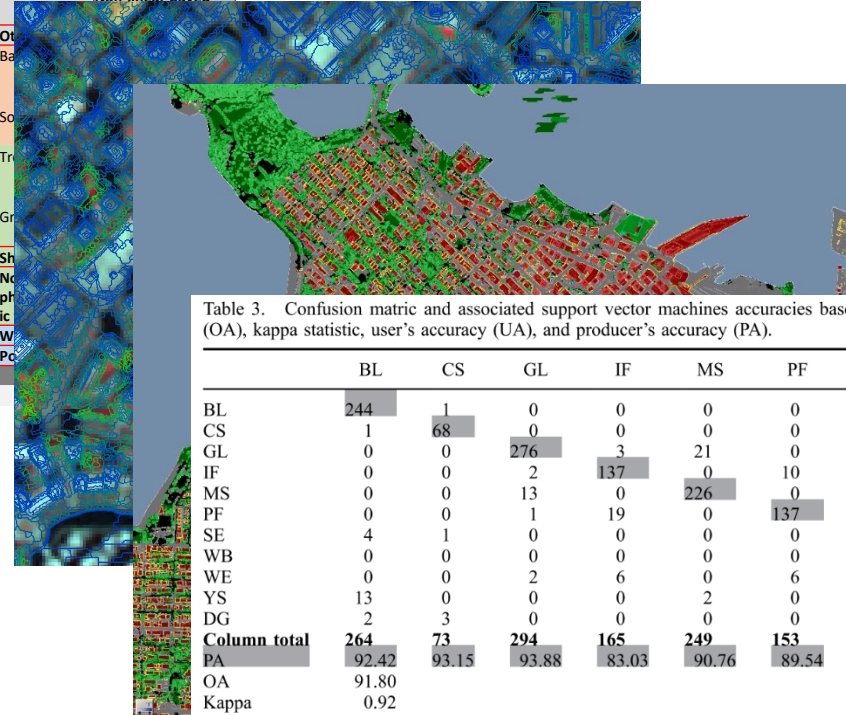


Table 3. Confusion matrix and associated support vector machines accuracies based on independence test data set. The accuracies include: overall accuracy (OA), kappa statistic, user's accuracy (UA), and producer's accuracy (PA).

	BL	CS	GL	IF	MS	PF	SE	WB	WE	YS	DG	Row total	UA
BL	244	1	0	0	0	0	1	1	0	8	1	256	95.31
CS	1	68	0	0	0	0	3	1	0	0	0	73	93.15
GL	0	0	276	3	21	0	0	0	1	0	0	301	91.69
IF	0	0	2	137	0	10	0	0	2	0	2	153	90.13
MS	0	0	13	0	226	0	0	0	1	2	1	243	93.00
PF	0	0	1	19	0	137	0	0	1	0	0	158	86.71
SE	4	1	0	0	0	0	79	0	0	0	1	85	92.94
WB	0	0	0	0	0	0	0	49	0	0	0	49	98.00
WE	0	0	2	6	0	6	0	0	167	3	0	184	90.76
YS	13	0	0	0	2	0	0	0	7	198	1	221	89.59
DG	2	3	0	0	0	0	1	0	0	0	76	82	92.68
Column total	264	73	294	165	249	153	84	51	179	211	82	1805	
PA	92.42	93.15	93.88	83.03	90.76	89.54	94.05	96.08	93.30	93.84	92.68		
OA	91.80												
Kappa	0.92												



Built-up	Buildings	
	Paved	Linear paved
		Linear paved elevated
		Non-linear paved features
	Other Built	
Bare	Barren	Natural barren
		Modified barren
		Linear unpaved
	Soil	Natural soil
		Modified soil
Vegetation	Tree canopy	Coniferous
		Deciduous
		Mixed
	Grass-herb	Modified grass-herb
		Natural grass-herb
	Shrub	
	Non-photosynthetic vegetation	
Water	Water	
	Pool	
Shadow		
Clouds/Ice		

MetroVan 2014 Class Definitions

- Classes informed by 2010 landcover classification and literature review
- 3-level hierarchy
- 21 “target” or final classes in total



Accuracy Assessment: Capturing Rare Cover Types



- Random points generated within 2010 Landcover layer
 - Stratified random sampling
- Points classified using Google Earth and aerial photos at 1m and 5 m
- Primary and Secondary class assignments

© Google Earth
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How do we deal with change ?

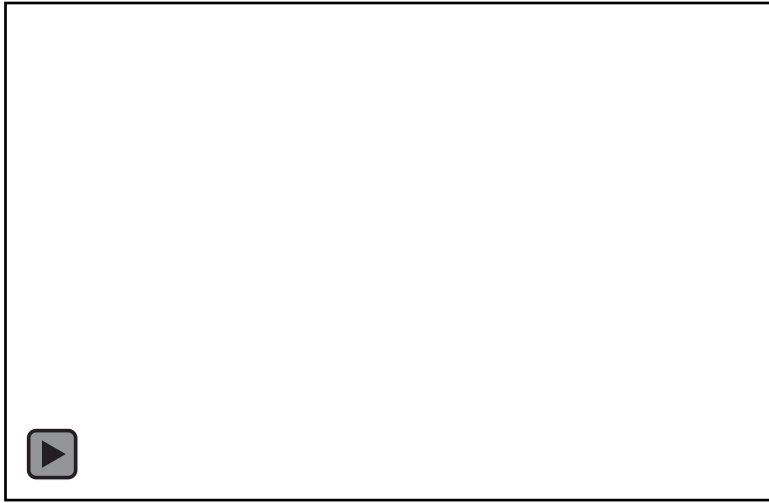
- Change on landscape is relatively small
- T2 – T1 approaches are very poor, when change is a small proportion of land base
 - i.e. 80% accuracy at T1 and T2 .. Yet change is 5%.
- Rather you have an accurate T1 and then monitor for pixels that have changed...
 - Maybe you can monitor using coarser "free" data at least to flag pixels as changed.



Take Home Message

- Need a high quality base image
 - Augment with LIDAR classes where available
 - Cost ?
- Need agreement on the desired classes / hierarchy
- Need to consider accuracy assessment and how to deal with land cover change





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