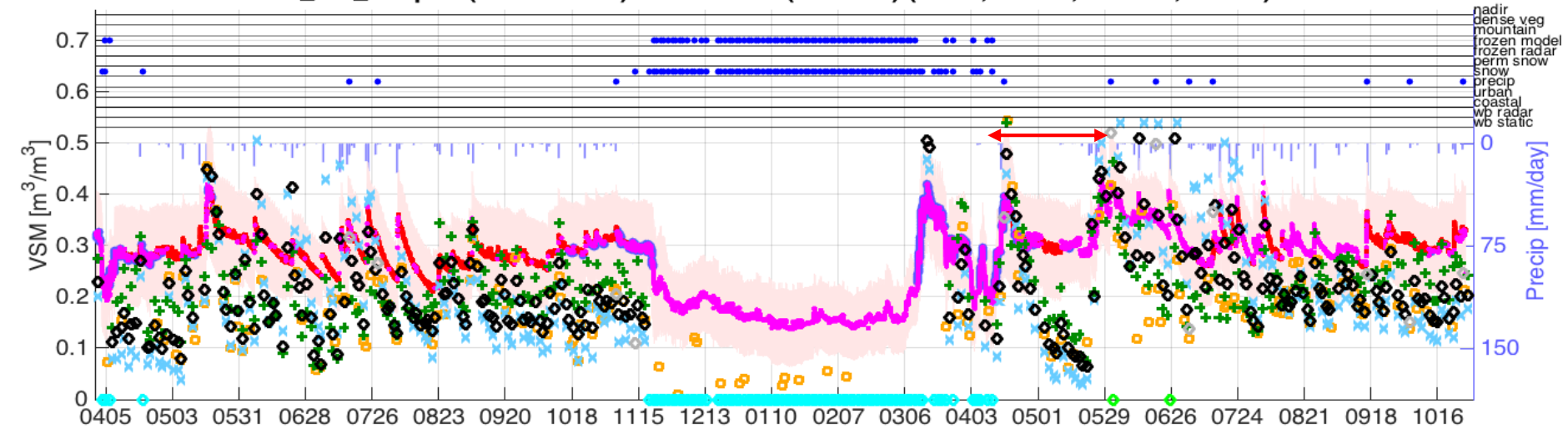


L2\_SM\_P-Opt 2 (R13080-999): 0901-36-01 (Carman) (49.58, -97.94; -2637.5, -578.5)



Alg	ubRMSE	Bias	RMSE	R
× SCA-H	0.092	-0.086	0.126	0.478
◆ SCA-V	0.058	-0.086	0.104	0.602
+ DCA	0.052	-0.076	0.092	0.519
□ SMOS	0.056	-0.108	0.122	0.729
• In Situ				

Climate class: Cold (Dfb)

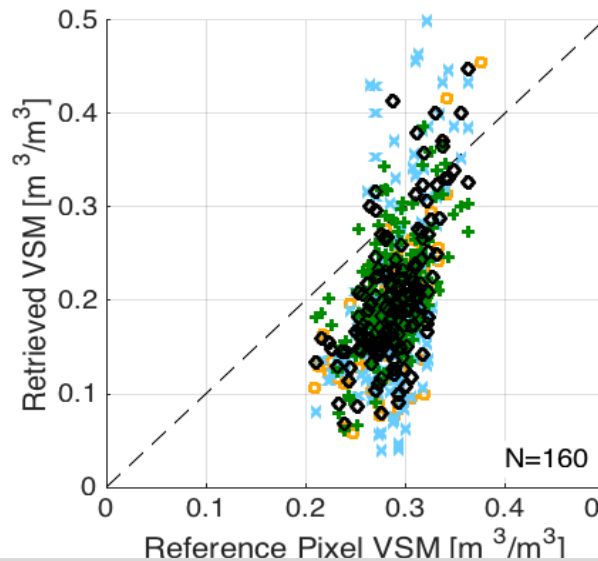
Landcover: Croplands

Soil texture:

S-%: 26

C-%: 34

BD: 1.18



- SMAP has an underestimation bias for all algorithms
- SMOS also has an underestimation bias and similar behavior
- During drydowns, SMAP decreases faster than the in situ obs
- The range of soil moisture is larger for SMAP
- The May 2016 time period is very interesting. Need further verification of what rainfall actually occurred during this period.

Black: Use recommended [Retrieval Quality Flag bit(0)=0]  
 Gray: Retrieval attempted and succeeded but use not recommended [bit(0)=1, bit(1)=0, bit(2)=0]  
 Green: Retrieval attempted but failed [bit(0)=1, bit(1)=0, bit(2)=1]  
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# Resolving Anomalous SMAP Retrievals: Carman

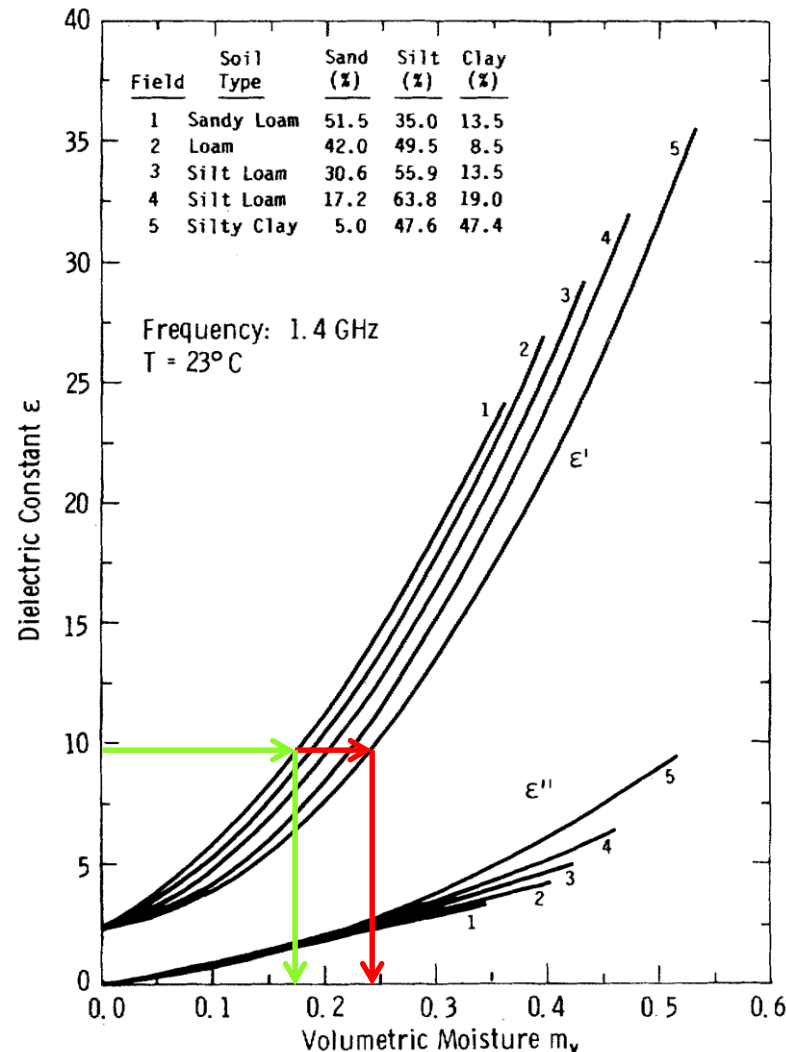
Source of Error	Resolve/Investigate	Expectation	Status of Assessment
Calibration of <i>in situ</i> probes	Conduct gravimetric/bulk density sampling and compare to probe estimates		Data were collected.
Bias in SMAP top-down versus <i>in situ</i> at depth	Profile sampling and vertical vs. horizontal probe measurements	Overestimation after rain, potential underestimation otherwise.	Need to compare the 0-5 to at 5 cm measurements collected. <i>Preliminary results from A. Colliander</i>
Soil texture	Sensitivity analysis and review dc vs. SM relationships	Clay soils can have lower dielectric constants for a given SM due to bound water. Underestimating clay fraction can lead to underestimating SM.	<i>Requested S. Chan to make some runs.</i>
Upscaling approach	Compare to a dense set of samples from field campaigns, etc. Compare alternative upscaling methods.		Comparison of RISMA network to ~45 temporary station (both vertical probes) showed a small underestimation bias for RISMA. <b>This does not explain the SMAP bias.</b>
Dielectric model	Review alternative methods	Different models produce varying SM For Hallikainen, underestimating the clay fraction underestimates SM	No action yet.
Soil emission model	All use Fresnel, which assumes uniform properties over the contributing depth		No action expected.
Surface roughness characterization	Do the conditions change significantly over time? Collect roughness data.	Underestimating roughness effects would underestimate SM	Data collected.
VWC (NDVI) climatology vs. actual	Compile data sets	Underestimating VWC would underestimate SM	A. Pacheco and R. Bindlish plots. Actual<estimate. <b>This does not explain the SMAP bias.</b>
b parameter	b values are based on limited experimental data sets		
Alternative tau estimates	Look at DCA and SMOS tau vs, climatology tau	Underestimating VWC would underestimate SM	<i>R. Bindlish 2015 plots, 2016 requested.</i>
Vegetation model	The current model is a simplification. However, it is difficult to implement more complex approaches Compare SMAP and SMOS parameters and retrievals		No action yet.
Spatial heterogeneity	Compare to PALS		Data were collected
RFI	Review data for potential issues	RFI would lead to underestimates of SM	No action expected.
Assuming Ts=Tv	Sensitivity analyses		No action yet.
Source of T data	Compare alternative products; ECMWF, Ka-band Is there a bias?		No action yet.
Temperature normalization model		Underestimating Ts will underestimate SM	No action yet.

# Based on J. Powers Slides

- If the RISMA network is calibrated and the upscaling is correct then the errors lie in the retrieval/validation process.
- Examine
  - Clay fraction and dielectric mixing model
  - Vertical vs. horizontal probes

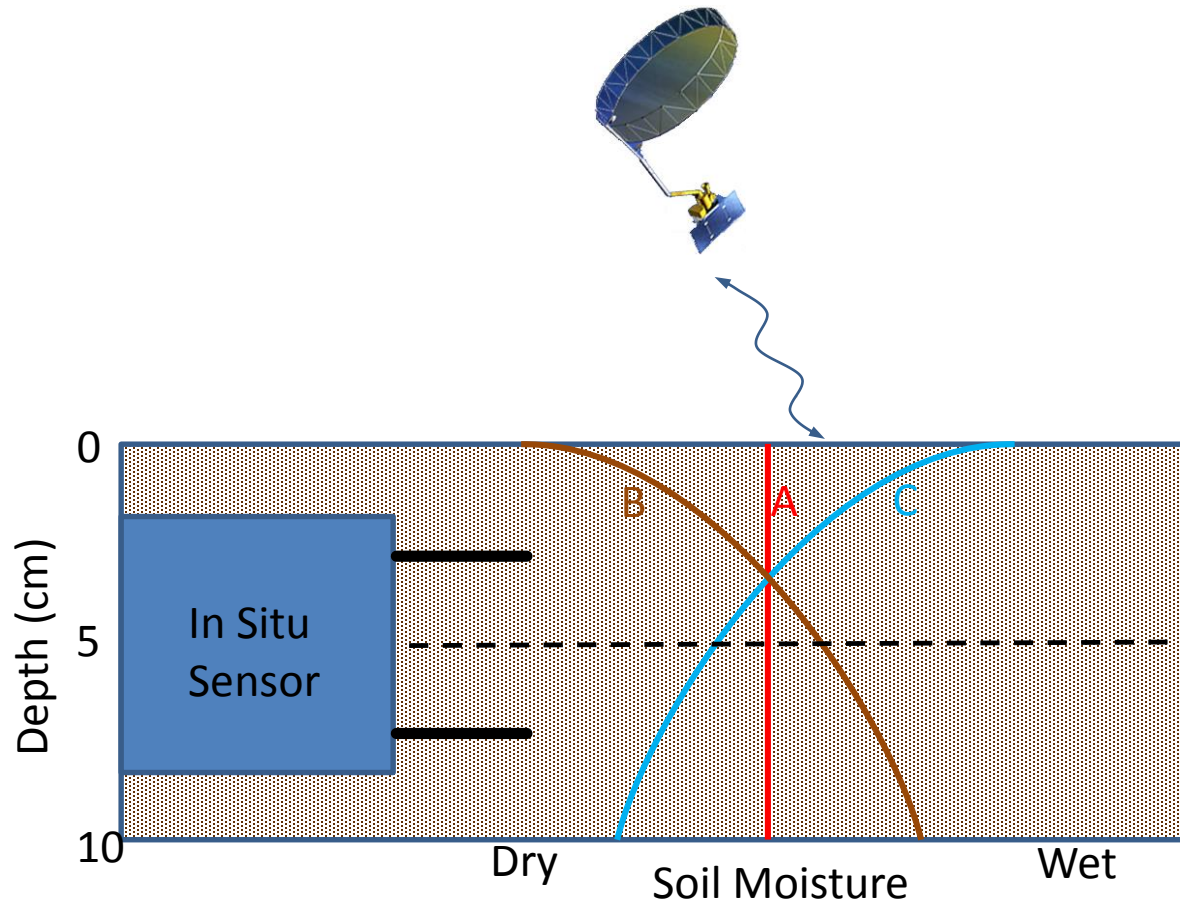
# Impact of Soil Texture on Dielectric Constant-Soil Moisture Relationships

- If the algorithm “retrieves” the dielectric constant and then uses the soil texture to estimate soil moisture, the soil moisture will be smaller for a sand than a clay.
- If there is an underestimation bias for soil moisture, decreasing the clay content would improve retrievals.



M. T. Hallikainen, F. T. Ulaby, M. C. Dobson, M. A. El-rayes, and L. Wu. 1985. Microwave Dielectric Behavior of Wet Soil-Part 1: Empirical Models and Experimental Observations. IEEE Transactions on Geoscience and Remote Sensing. GE-23: 25-34, DOI: 0.1109/TGRS.1985.289497

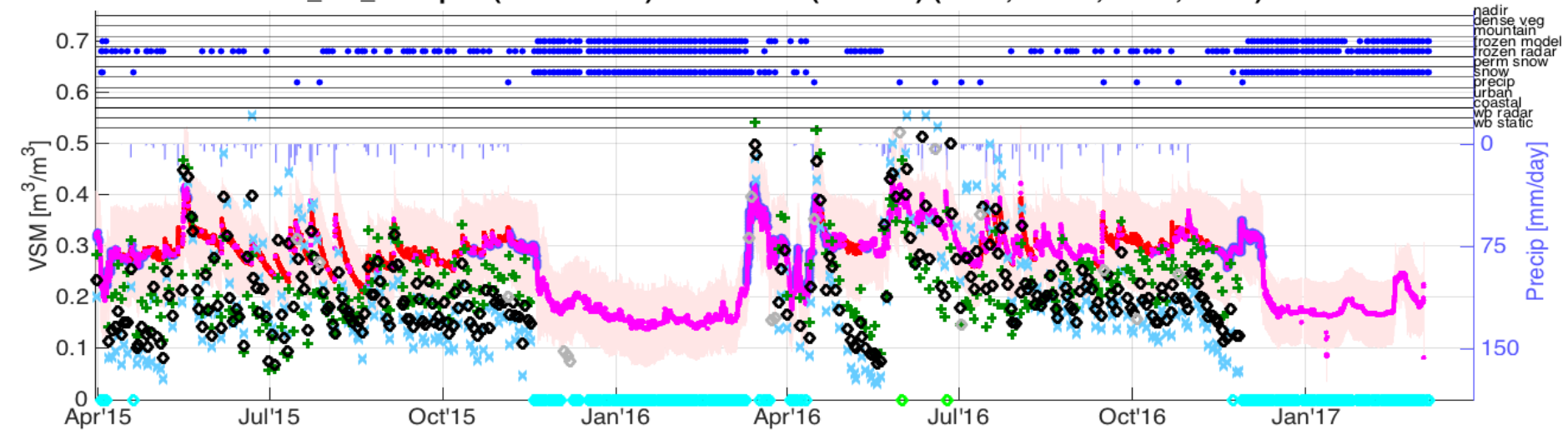
# Top-down Versus at Depth Measurements



- A. Uniform soil moisture with depth: OK!
- B. Dry down: Satellite may underestimate vs. in situ
- C. Recent rain/irrigation: Satellite may overestimate vs. in situ

# Carman (Core Pixel)

L2\_SM\_PE-Opt 2 (R14010-001): 0901-33-01 (Carman) (49.60, -97.98; 879.0, 193.0)



Alg	ubRMSE	Bias	RMSE	R
× SCA-H	0.092	-0.092	0.130	0.434
◊ SCA-V	0.055	-0.090	0.105	0.599
+ DCA	0.054	-0.073	0.091	0.566
• In Situ				

Climate class: Cold (Dfb)

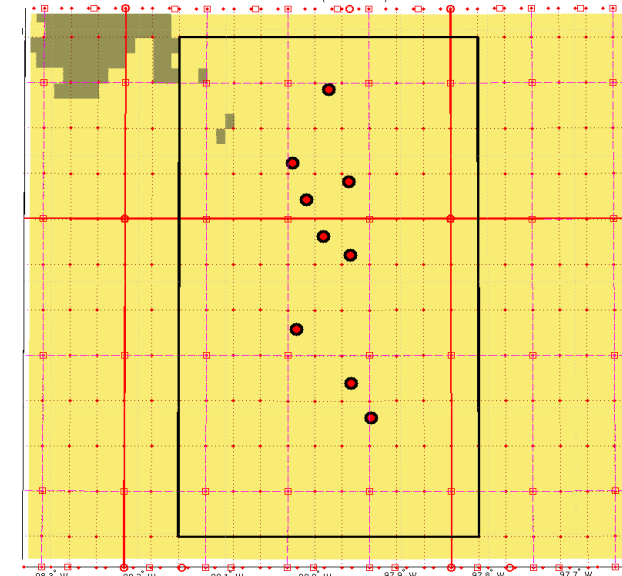
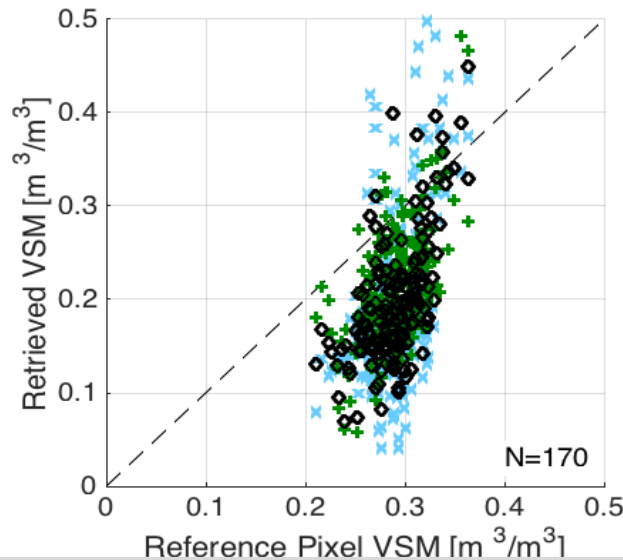
Landcover: Croplands

Soil texture:

S-%: 30

C-%: 31

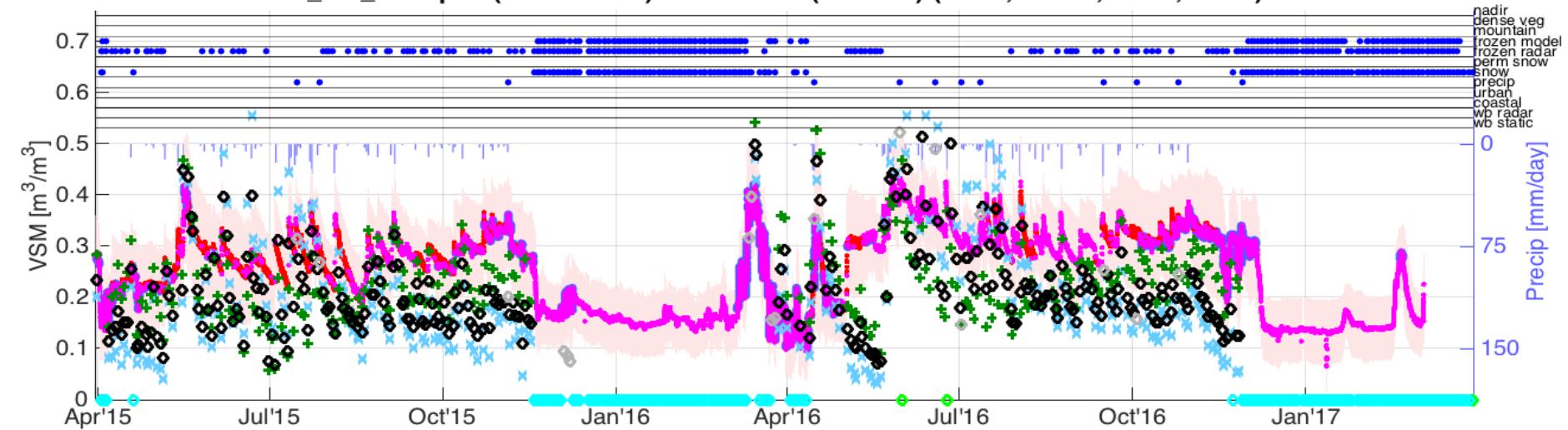
BD: 1.18



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# Carman (Candidate Pixel)

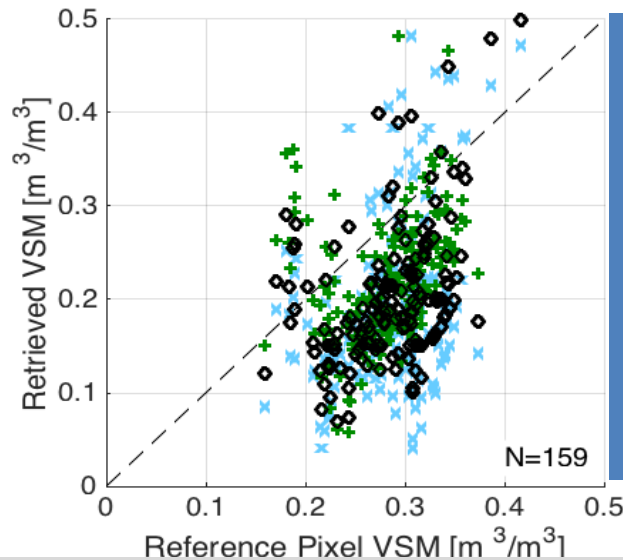
L2\_SM\_PE-Opt 2 (R14010-001): 0901-33-02 (Carman) (49.60, -97.98; 879.0, 193.0)



Alg	ubRMSE	Bias	RMSE	R
SCA-H	0.095	-0.086	0.128	0.324
SCA-V	0.068	-0.081	0.106	0.416
DCA	0.071	-0.061	0.094	0.357
In Situ				

Climate class: Cold (Dfb)  
Landcover: Croplands

Soil texture:  
S-%: 30  
C-%: 31  
BD: 1.18



- The range of in situ soil moisture increases when we use the vertical probe data (as opposed to the horizontal).
- Bias improves a bit but all other metrics degrade.
- Results might support using a shallower depth for validation.

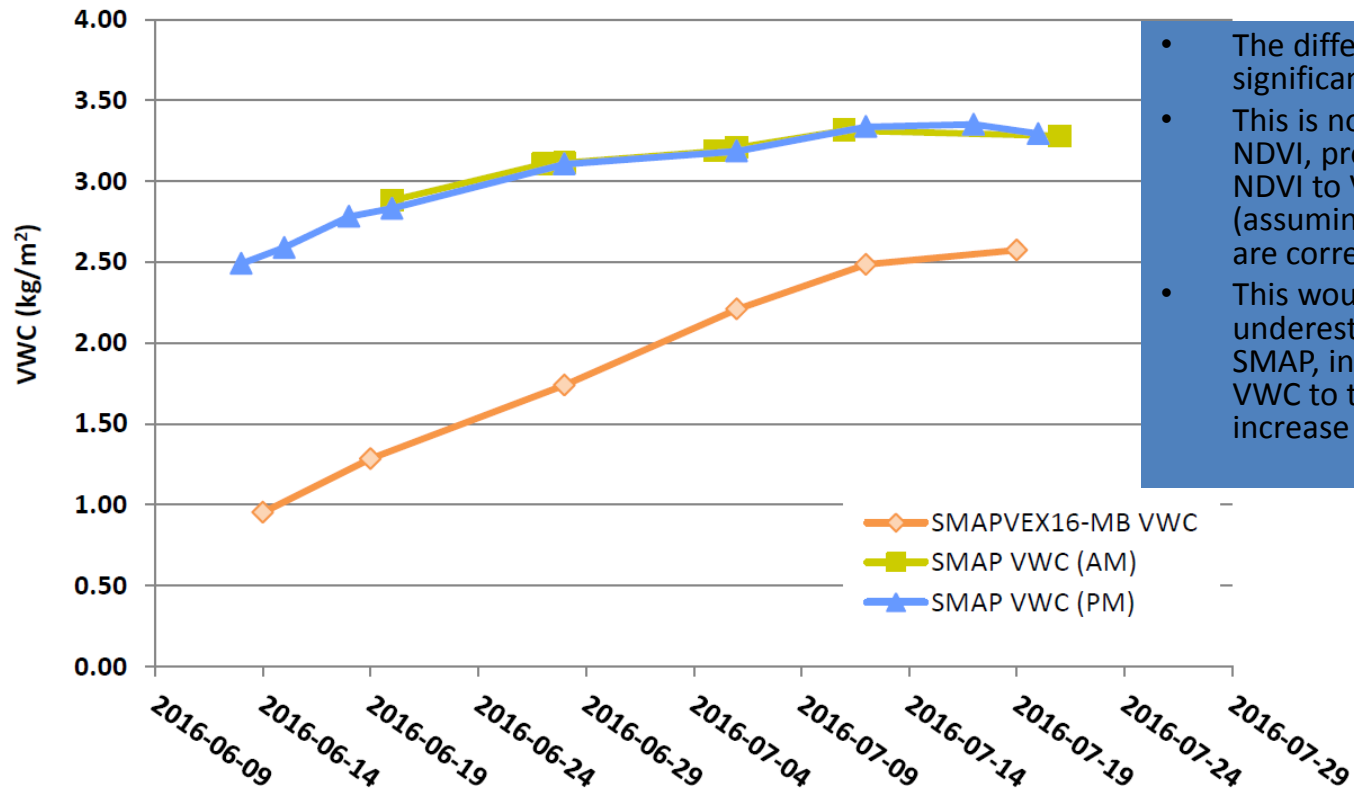
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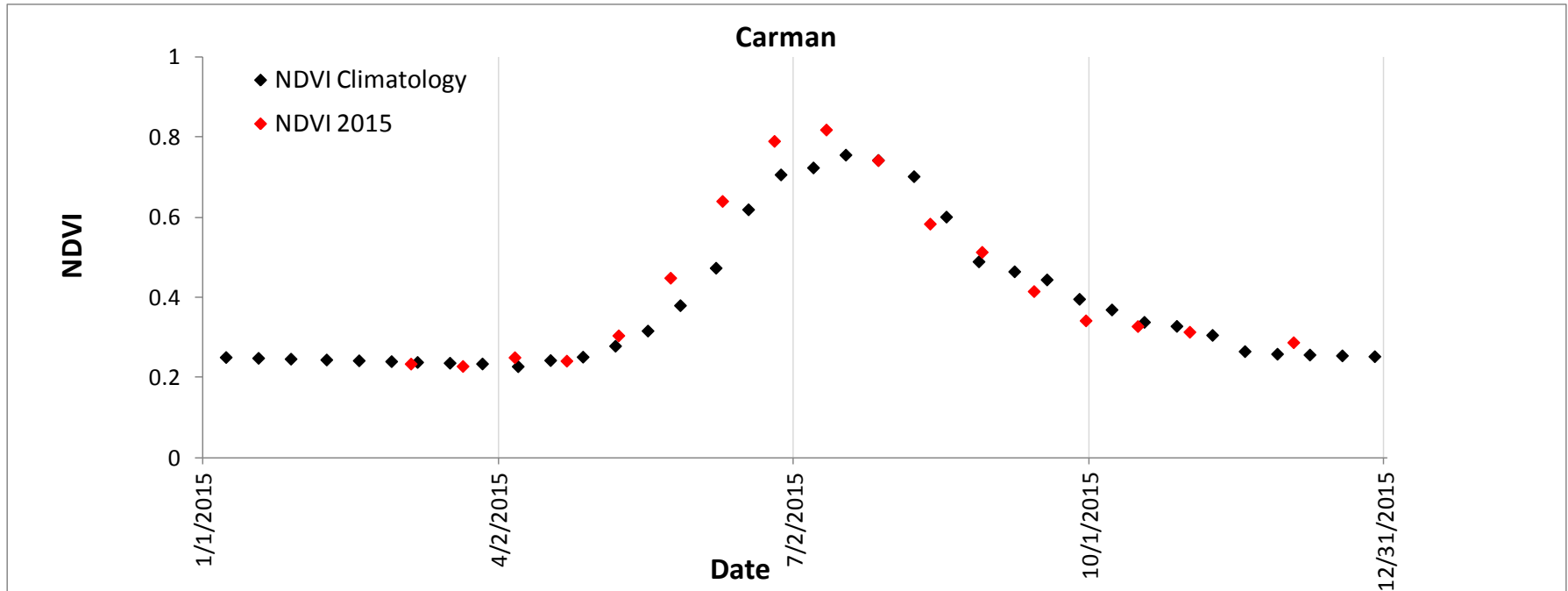


# Comparison of SMAPVEX16-MB VWC and SMAP Climatology-based Estimates (From Anna Pacheco)

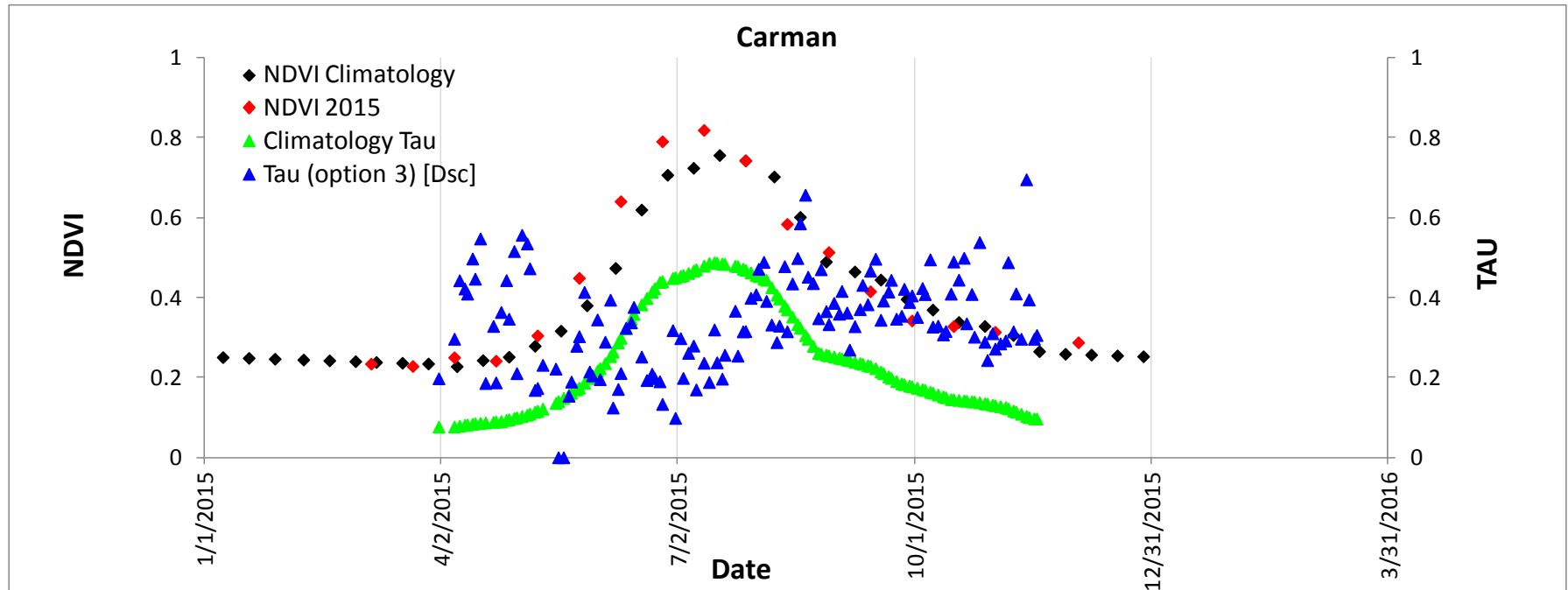


- The differences in VWC are significant.
- This is not associated with NDVI, probably need a better NDVI to VWC function (assuming campaign results are correct).
- This would not explain underestimation bias of SMAP, in fact decreasing VWC to these levels would increase the bias.

# Comparison of Actual and NDVI Climatology for 2015 (Carman)



# Comparison of Actual and NDVI and Tau Climatology for 2015 (Carman)



- What's happening with the DCA retrieved VWC???

# What Would “fix” the Carman Soil Moisture Retrievals

- Increase surface roughness
- Increase tau (~~VWC/NDVI~~ or b) (Omega?)
- Use a surface biased in situ observation or shallower depth
- Decrease clay fraction
- Decrease physical temperature