Exploring the Ameriflux Data

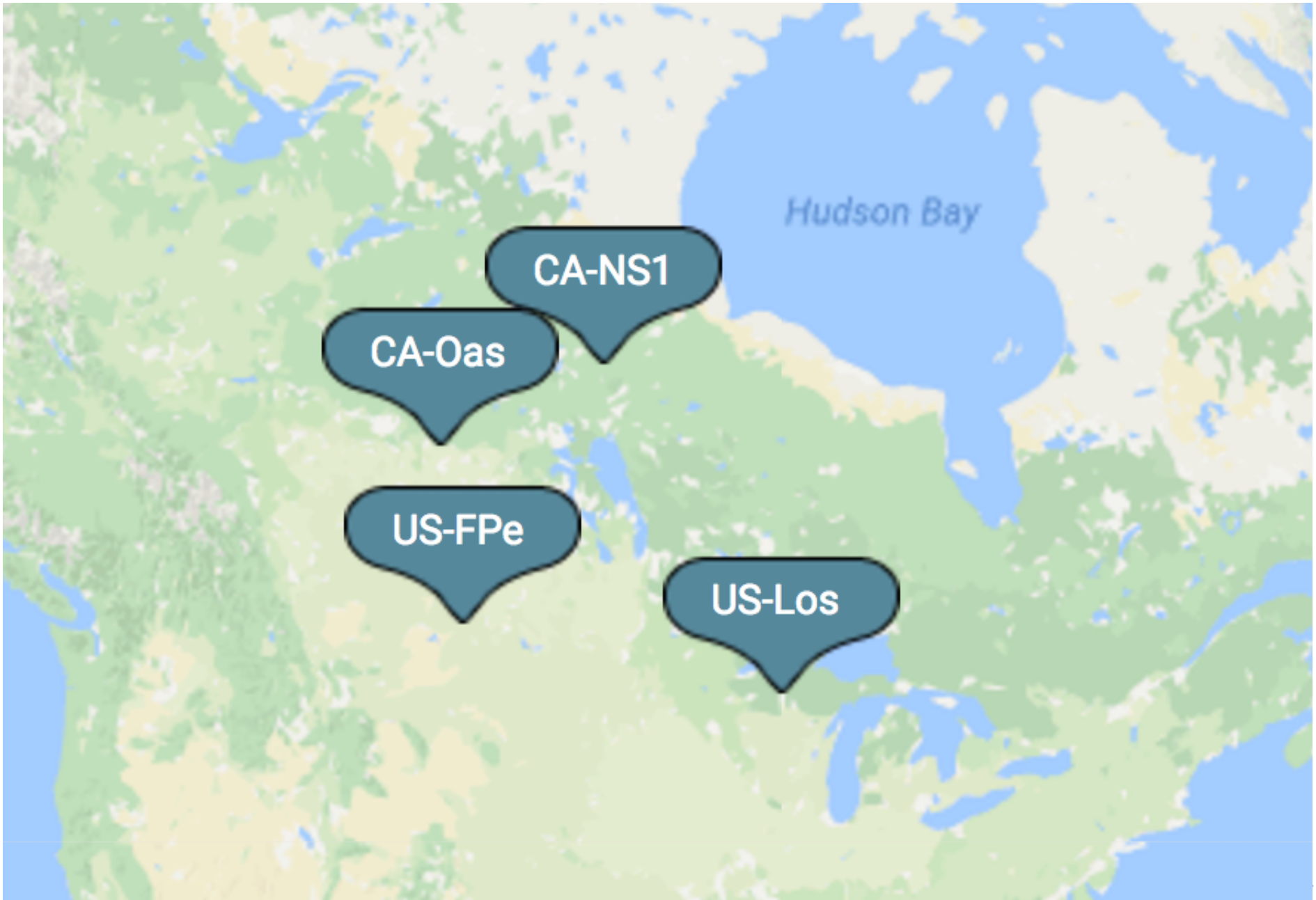
Trends and Relationships

I. Markelov

Upscaling Workgroup Meeting

June 16, 2017

Sites

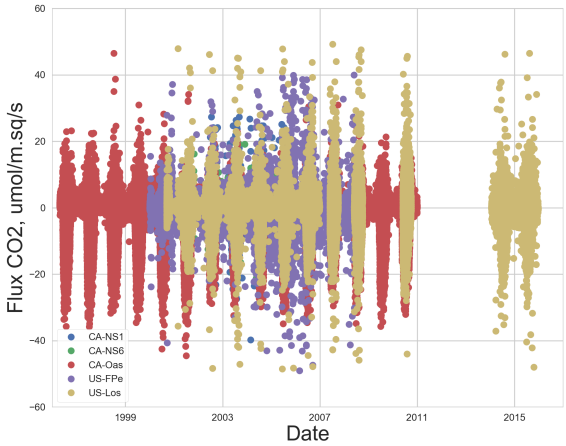


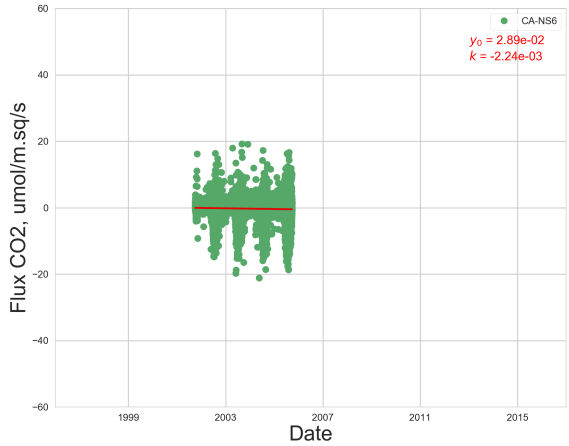
Sites

Name	Climate	Type	Mean T, C	Mean P, mm
CA-NS1	Dfc	ENF	-2.89	500.29
CA-NS6	Dfc	OSH	-3.08	495.37
CA-Oas	Dfc	DBF	0.34	428.53
US-FPe	Bsk	GRA	5.48	334.8
US-Los	Dfb	WET	4.08	828

Climate: Dfc - Subarctic: severe winter, no dry season, cool summer; Bsk - Cold semi-arid climate, steppe, warm winter; Dfb - Warm Summer Continental: significant precipitation in all seasons.
Vegetation: ENF - Evergreen Needleleaf Forests; OSH - Open Shrublands; DBF - Deciduous Broadleaf Forests; GRA - grassland; WET - Permanent Wetlands.

Flux in Time


Flux CO2, umol/m.sq/s
Date
CA-NS1
CA-NS6
CA-Oas
US-FPe
US-Los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CA-NS6
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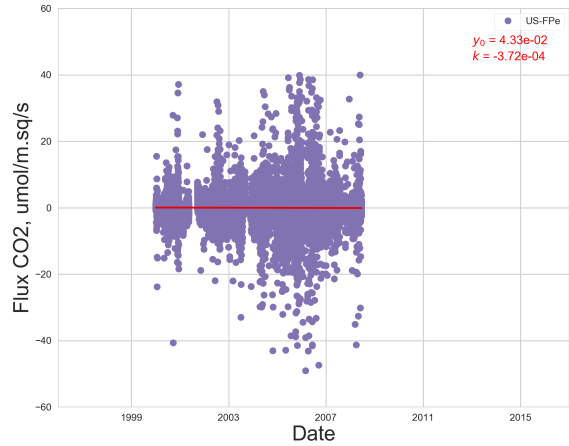
Flux CO2, umol/m.sq/s
Date
US-FPe
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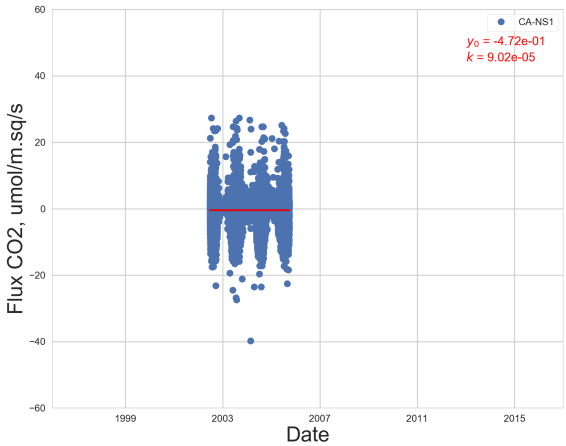
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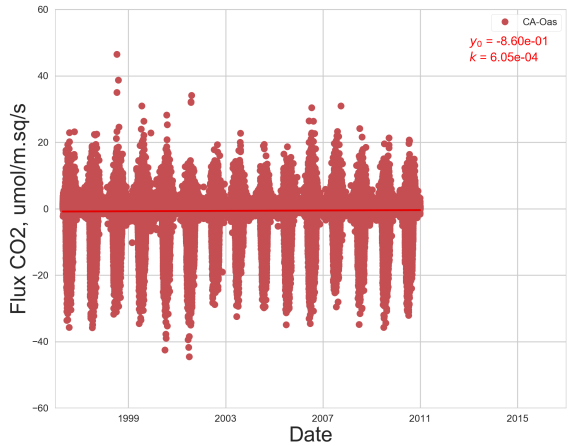
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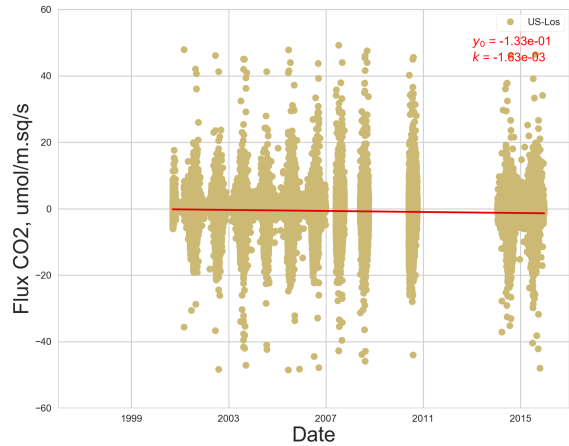
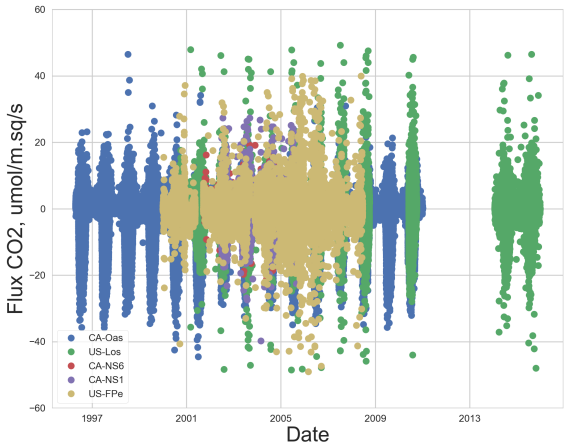
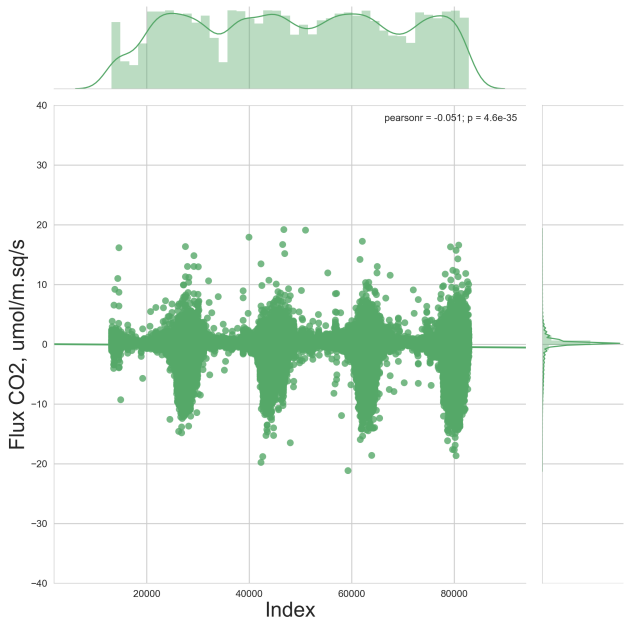
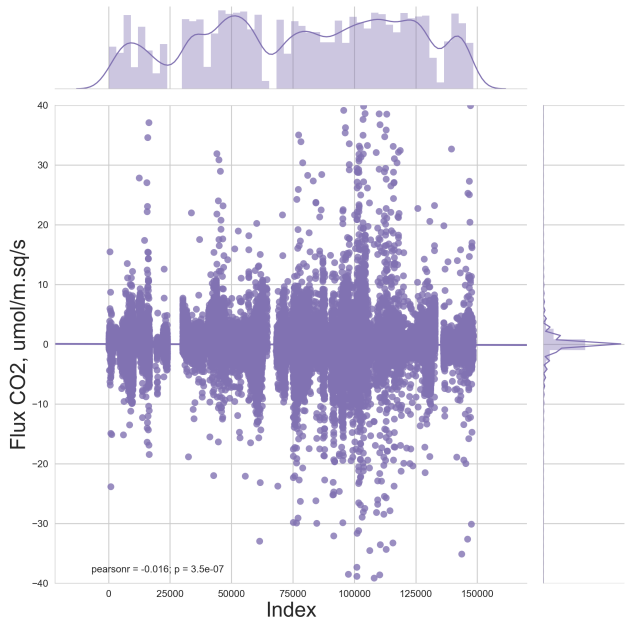
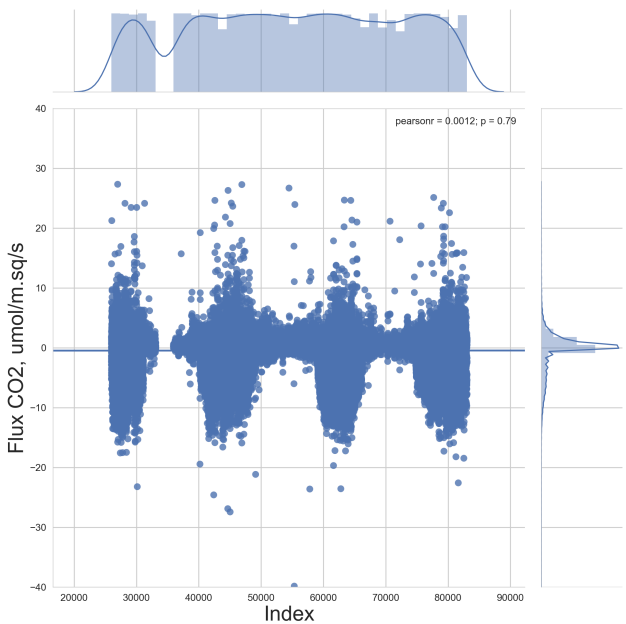

Flux CO2, umol/m.sq/s
Date
US-Los
y_0 = -1.33e-01

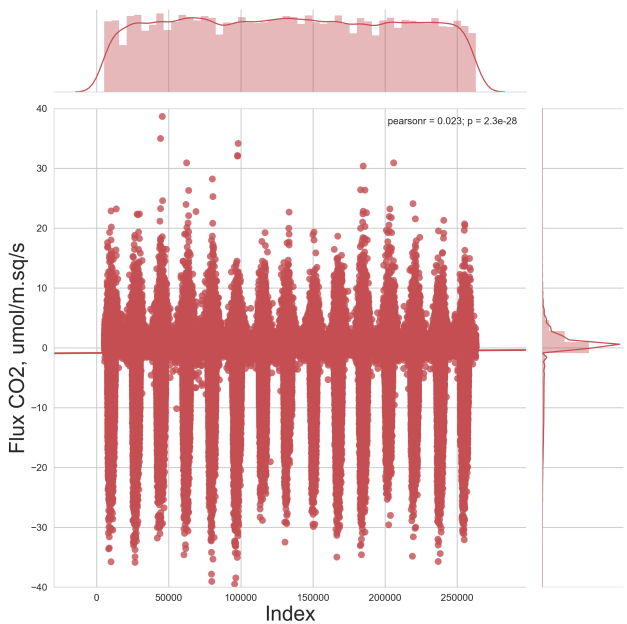
Density of Measurements

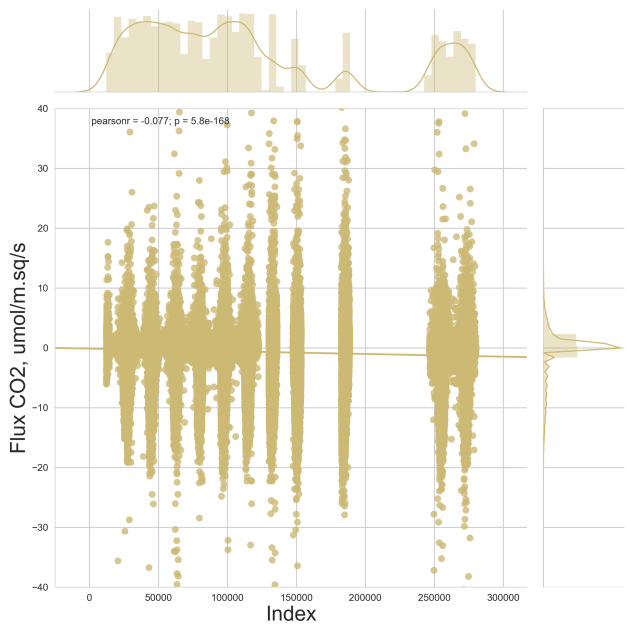




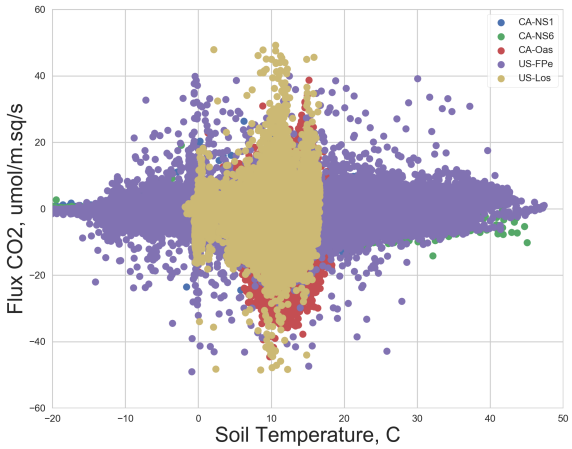


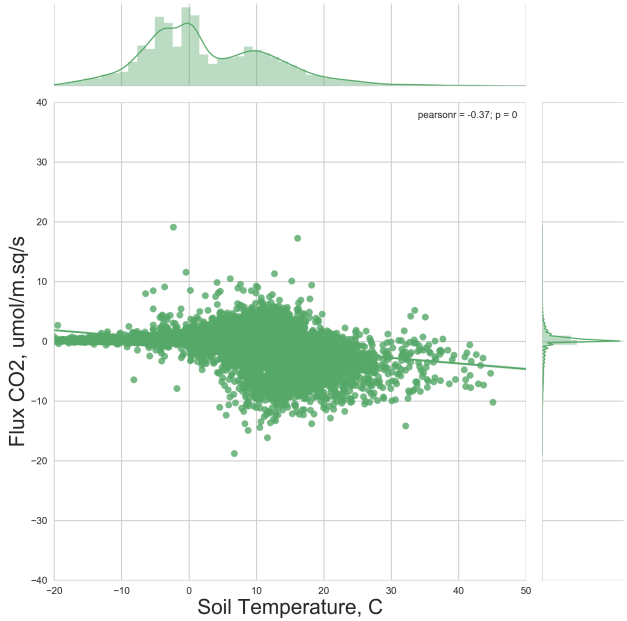


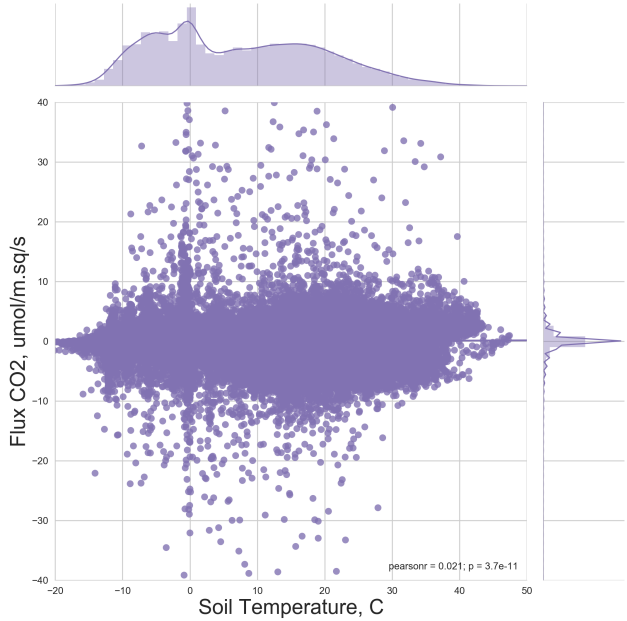


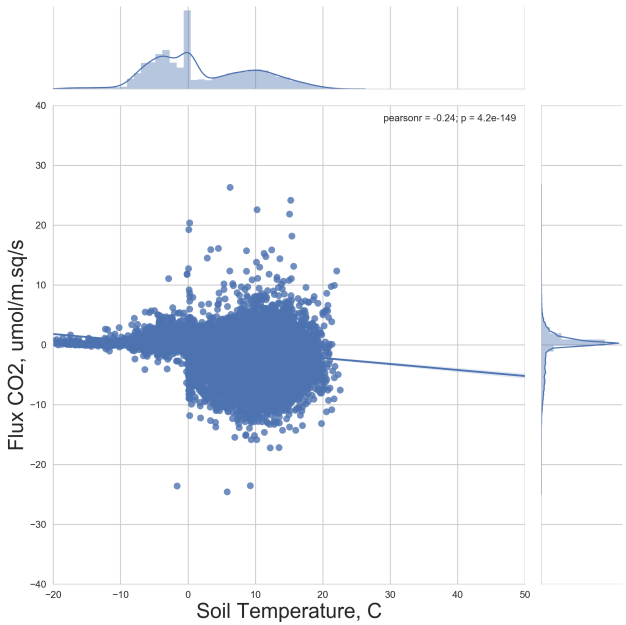


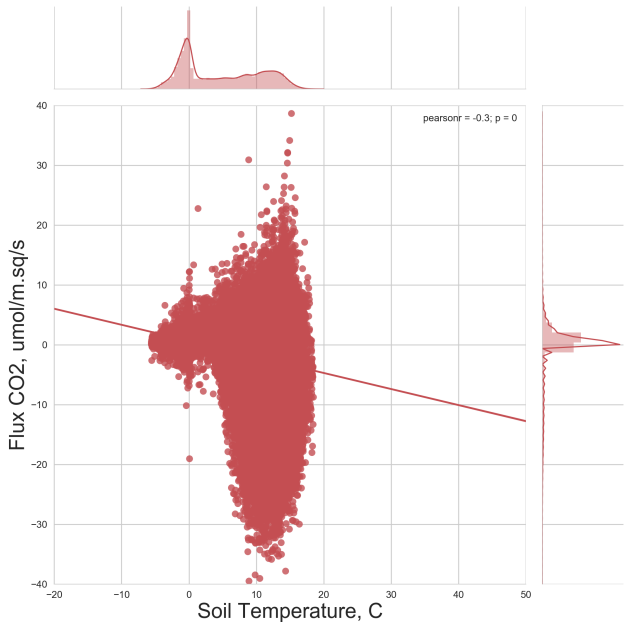
Flux vs Soil Temperature

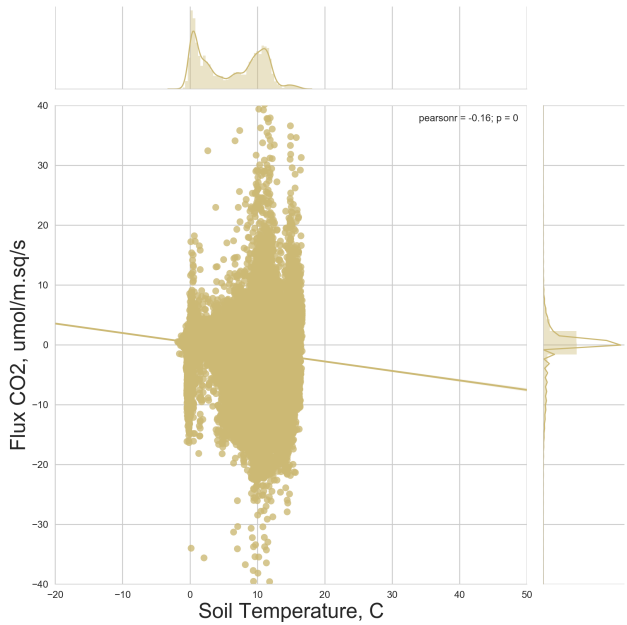




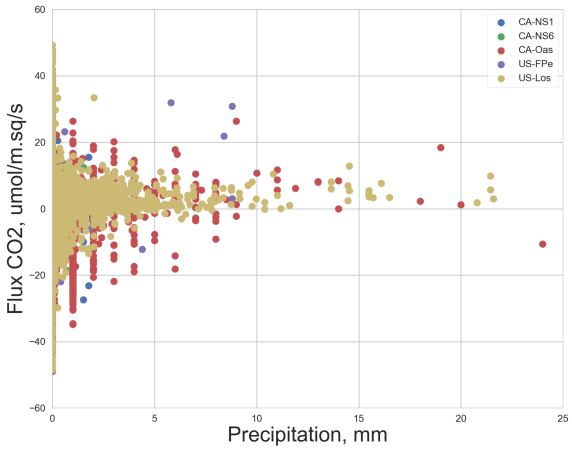


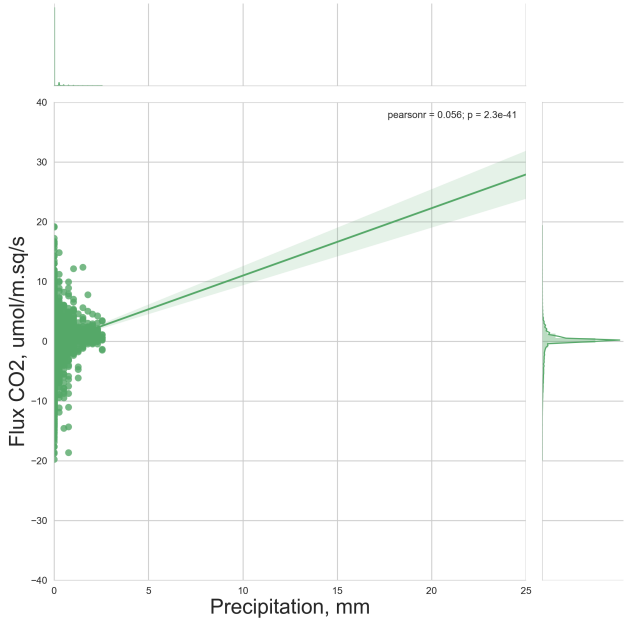


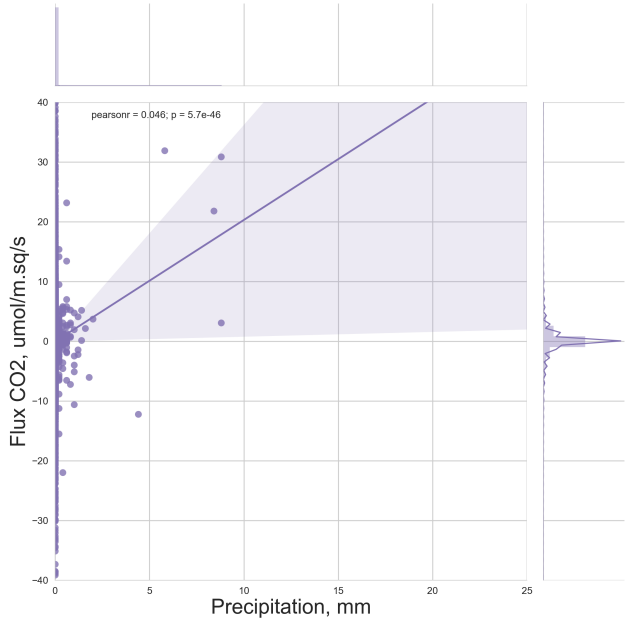


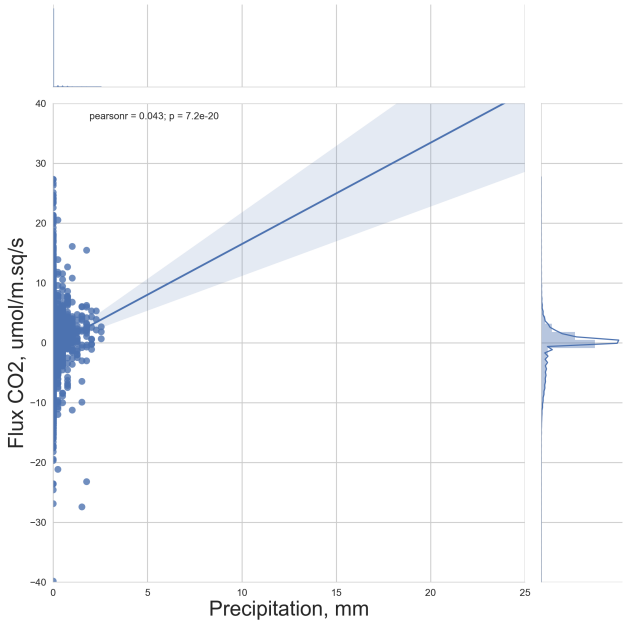


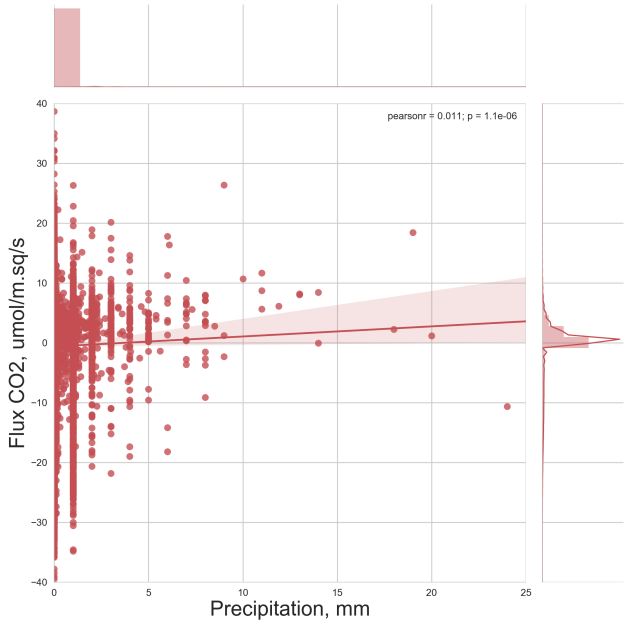
Flux vs Precipitation

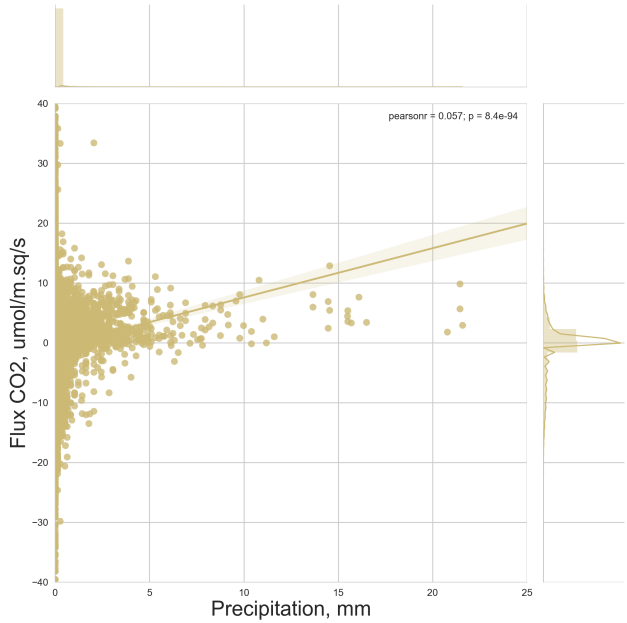




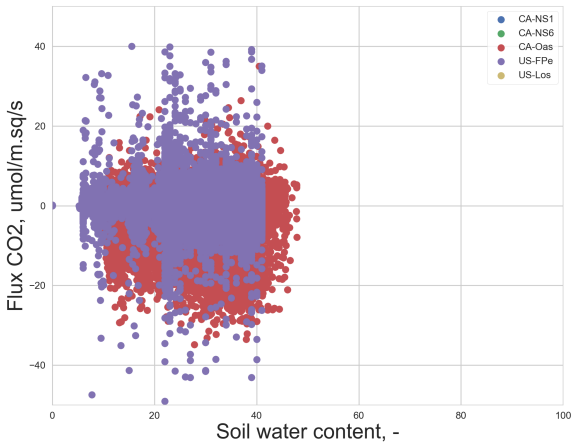


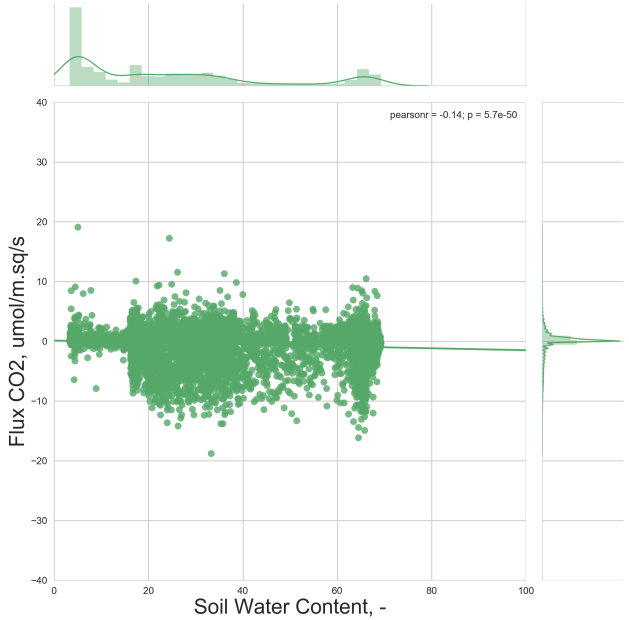


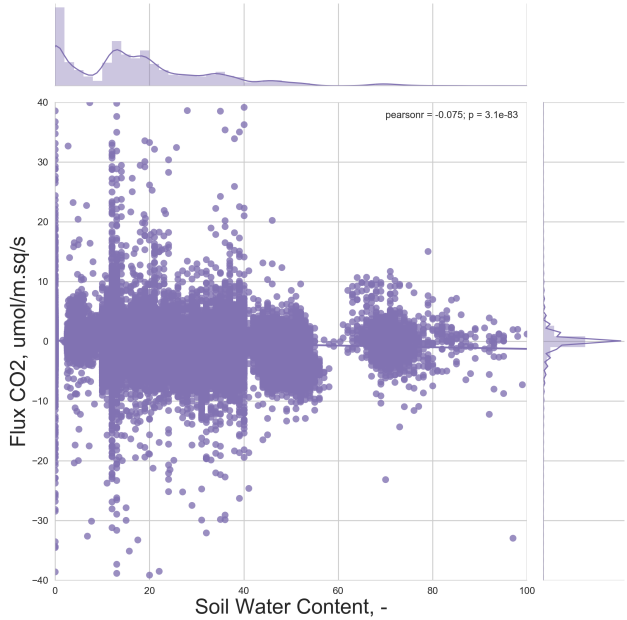


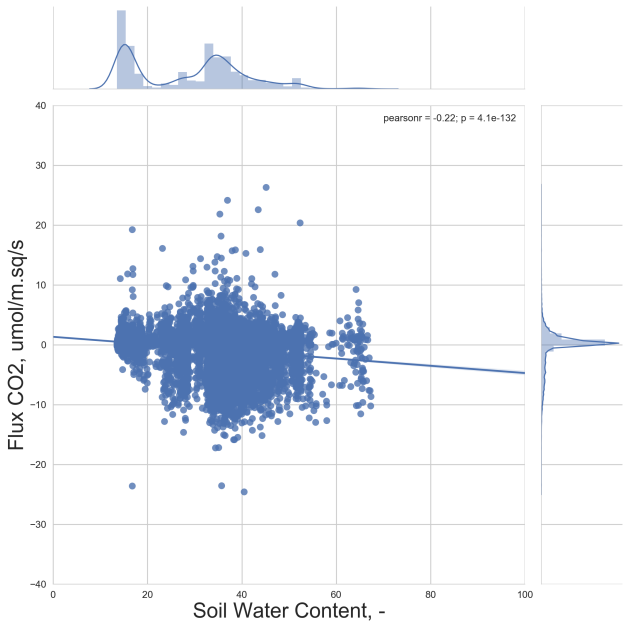


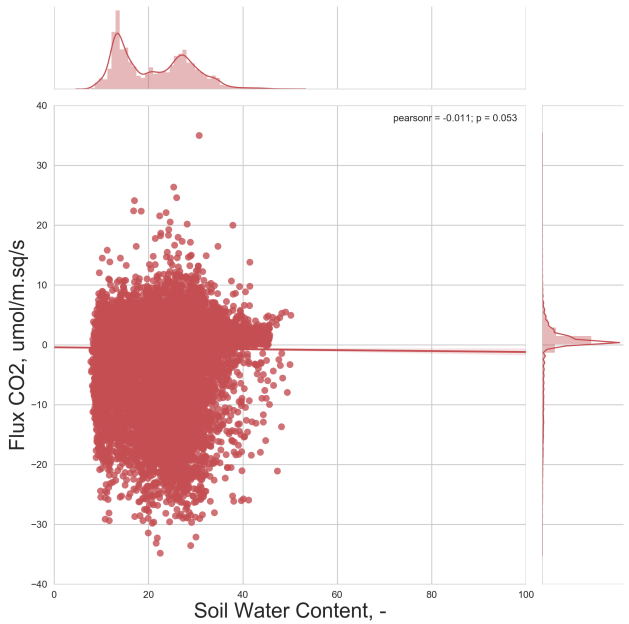
Flux vs Soil Water Content

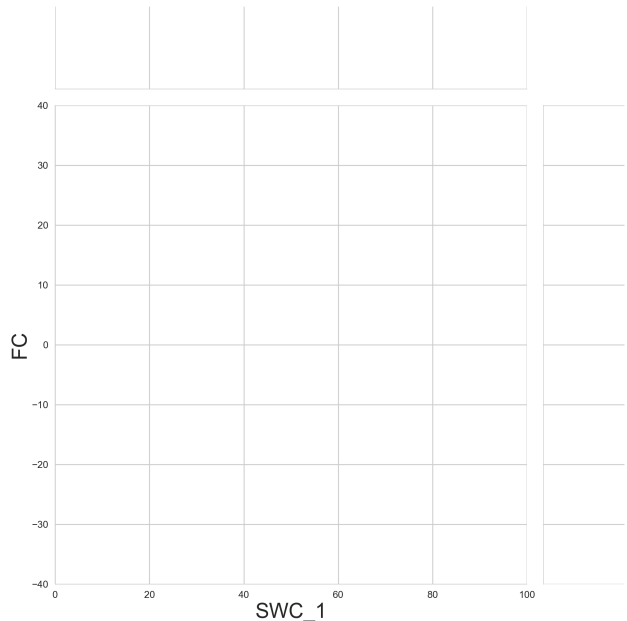




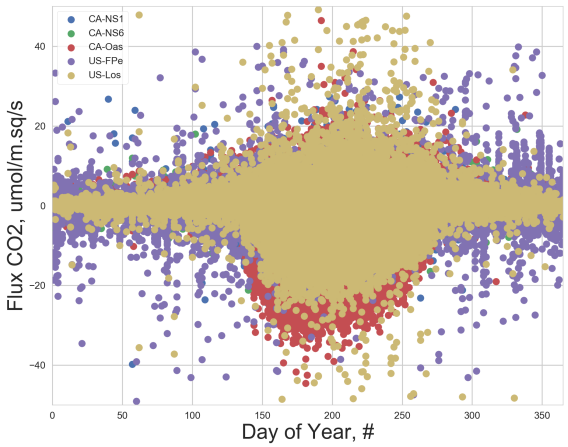


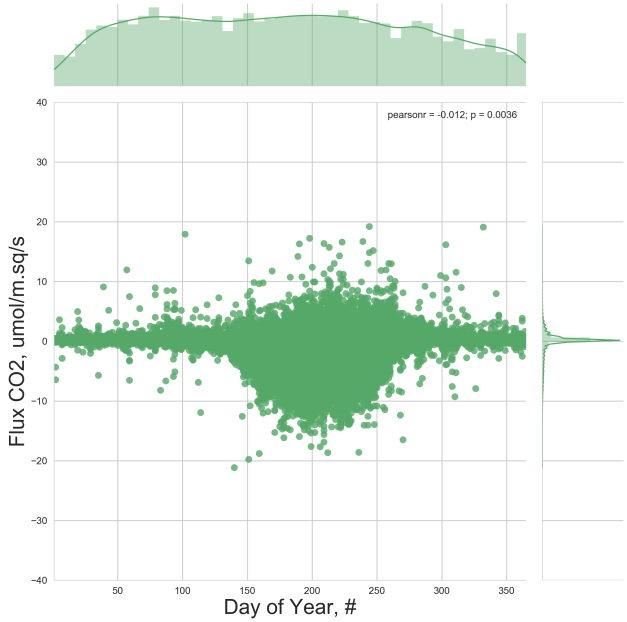


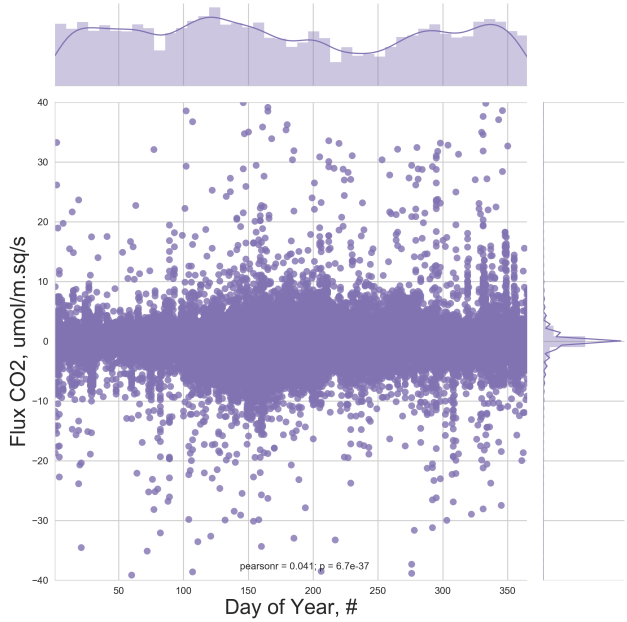


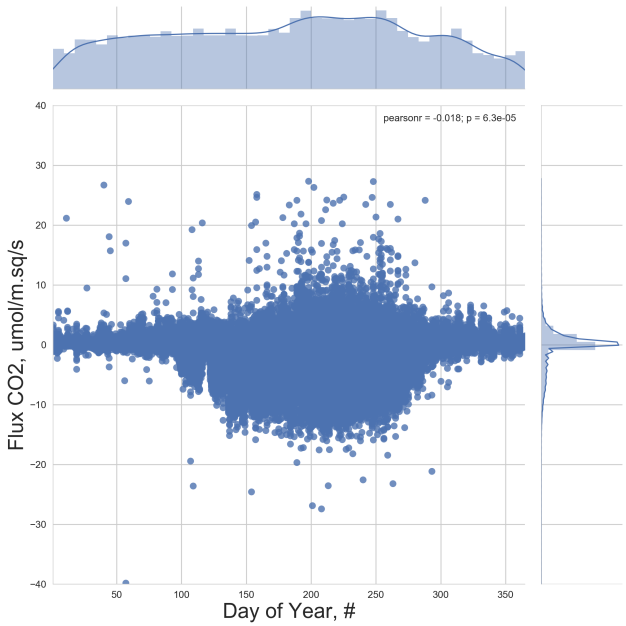


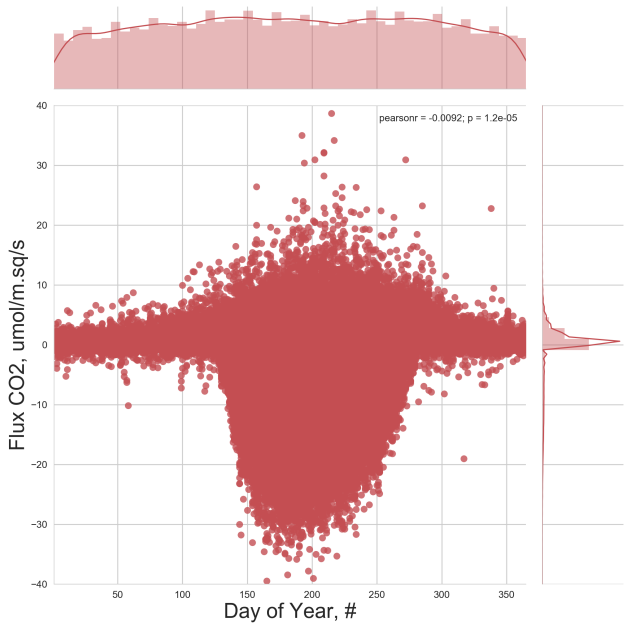
Flux on Yearly scale

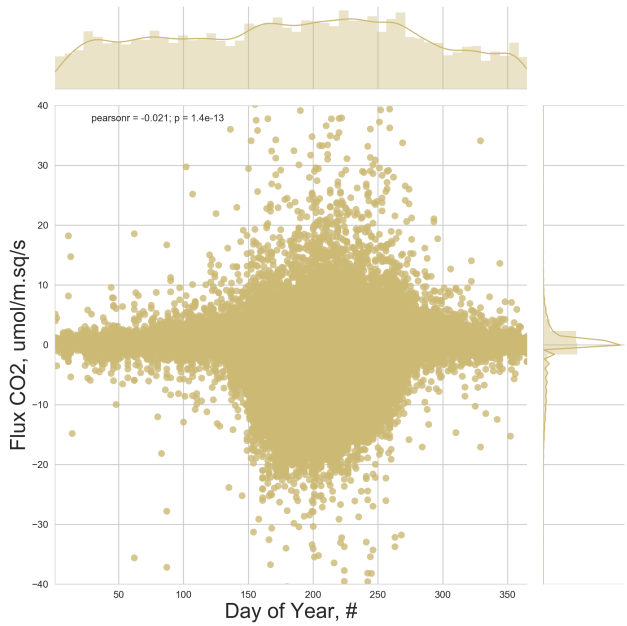




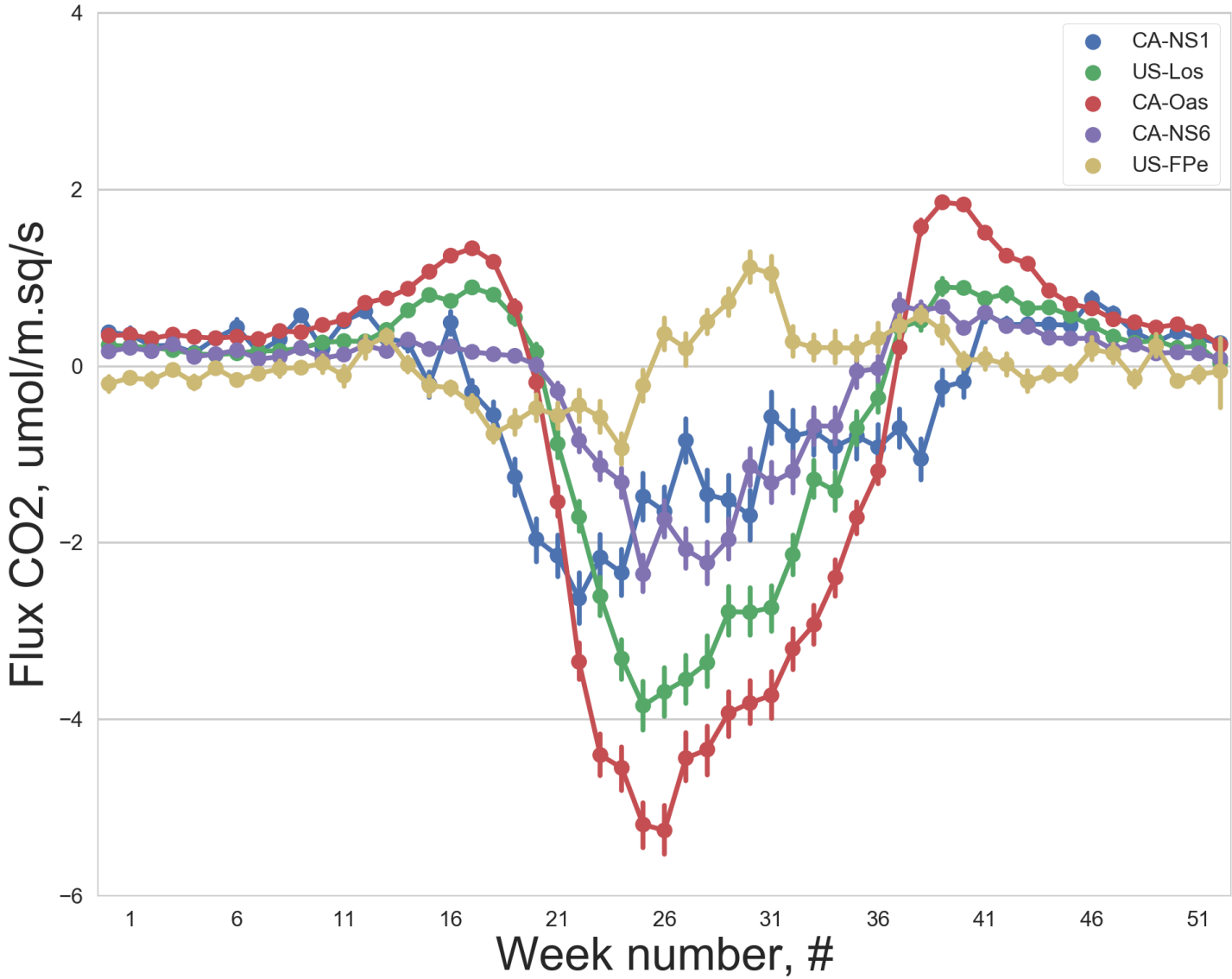




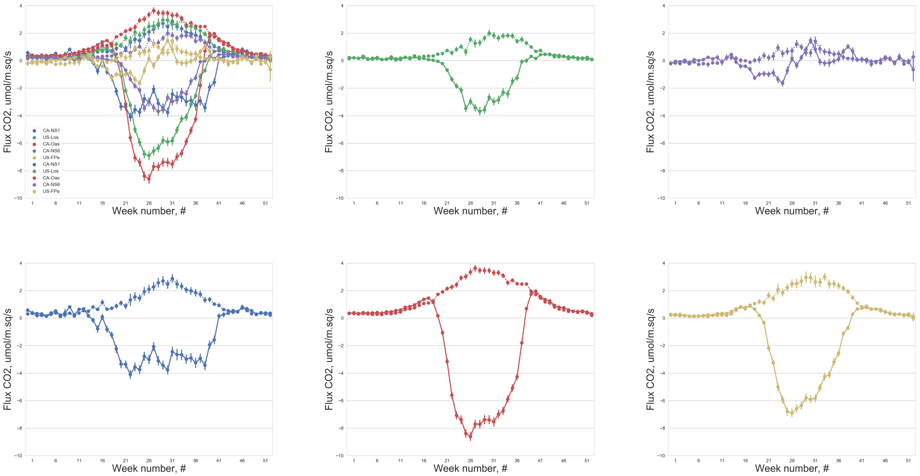




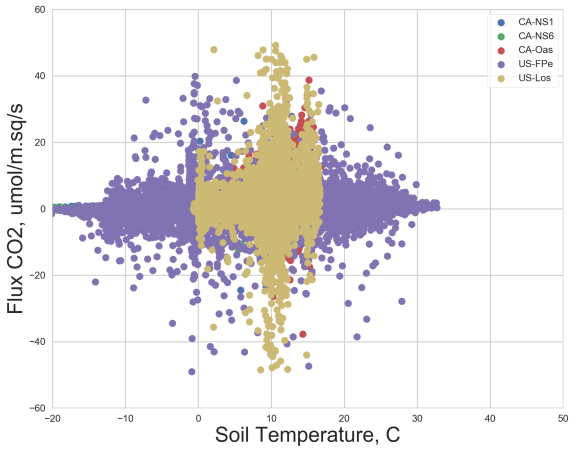
Weekly average plots

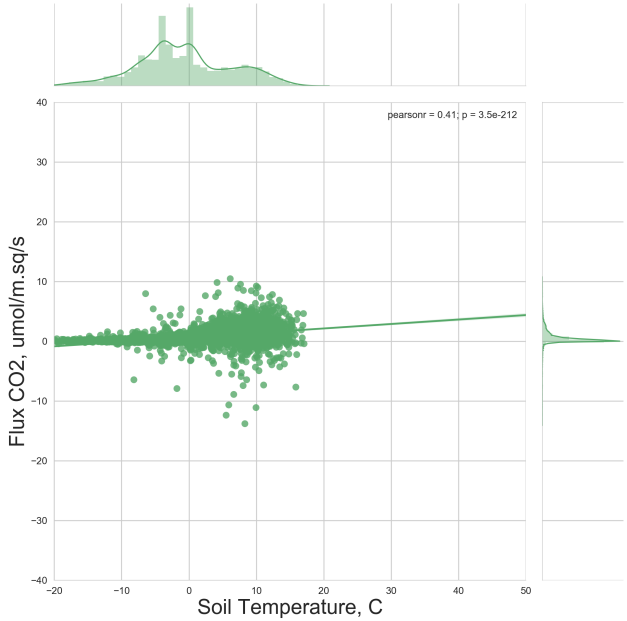


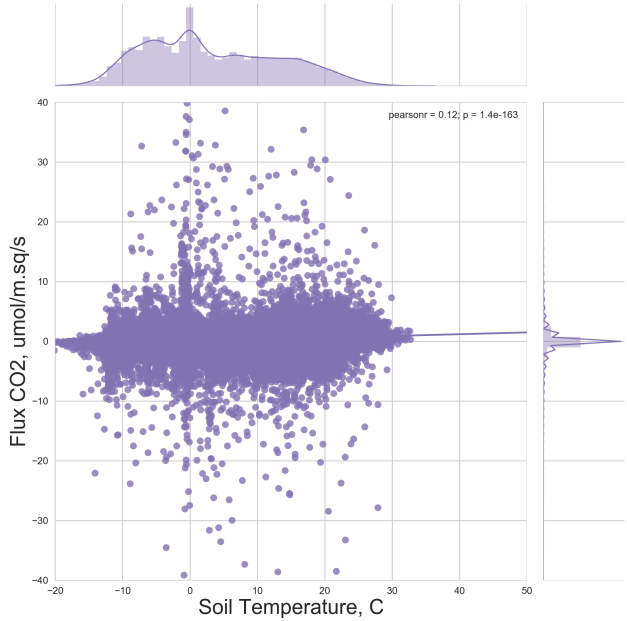
Weekly average: Day/Night

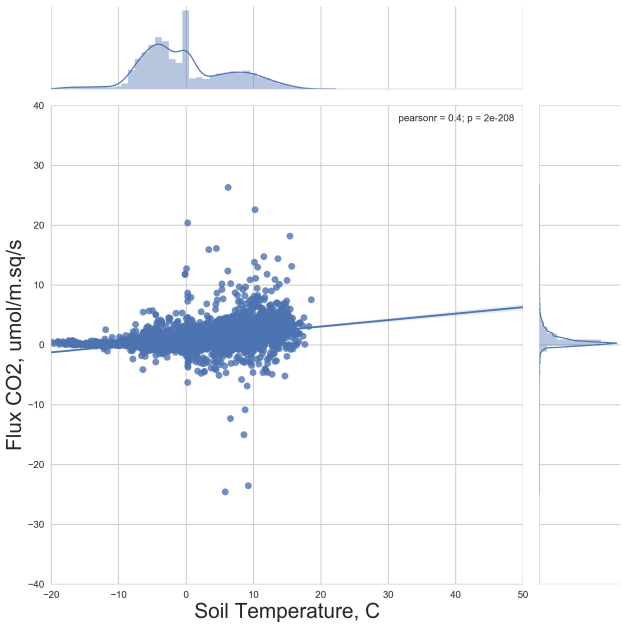


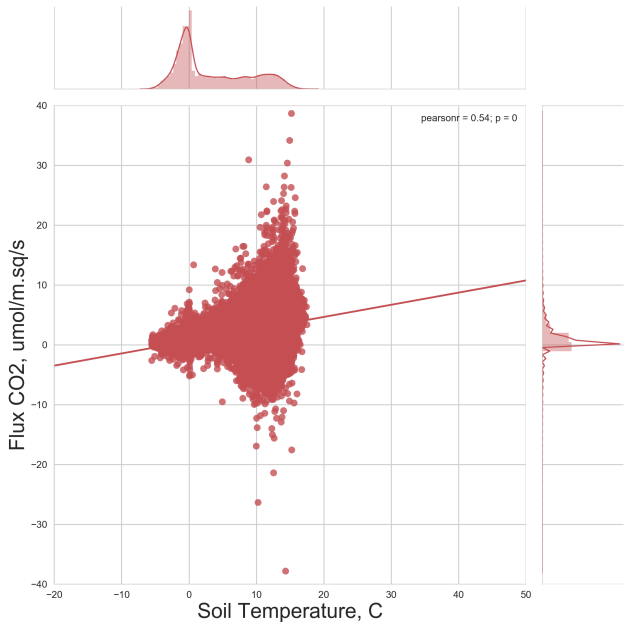
Flux vs Temperature at Night

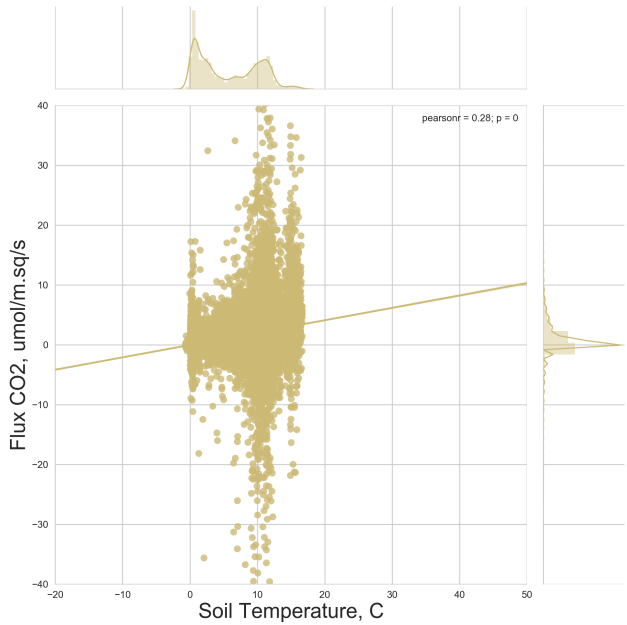




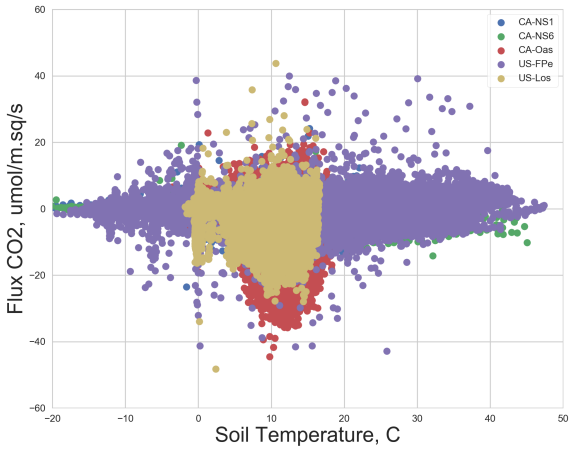


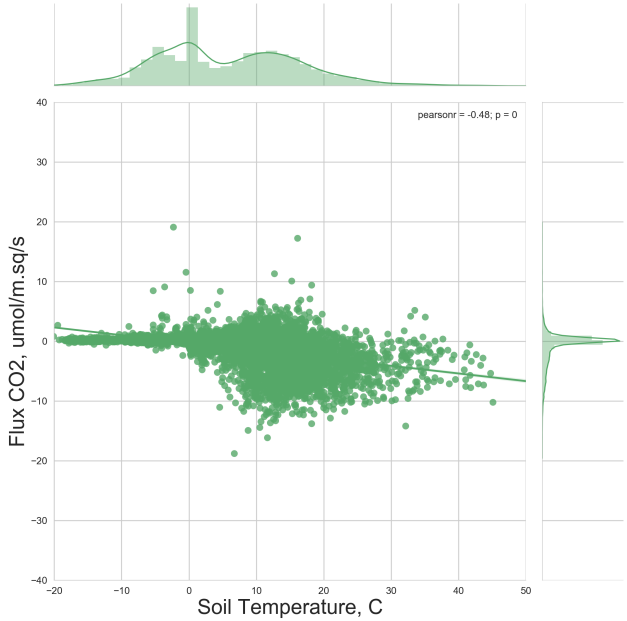


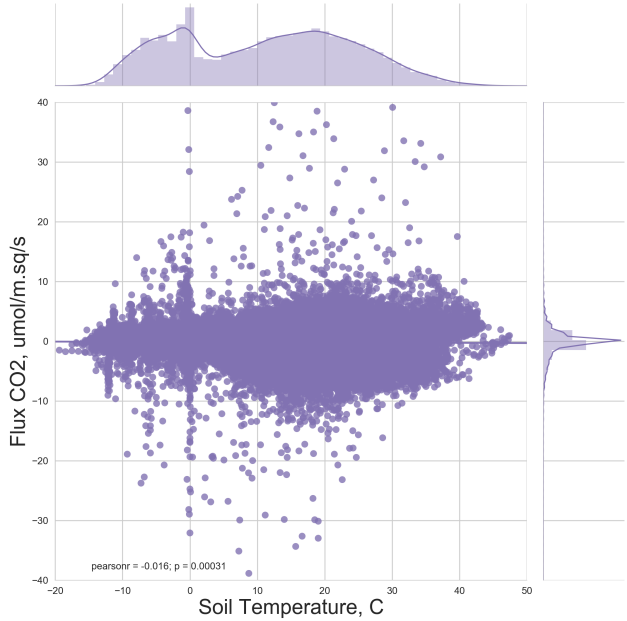


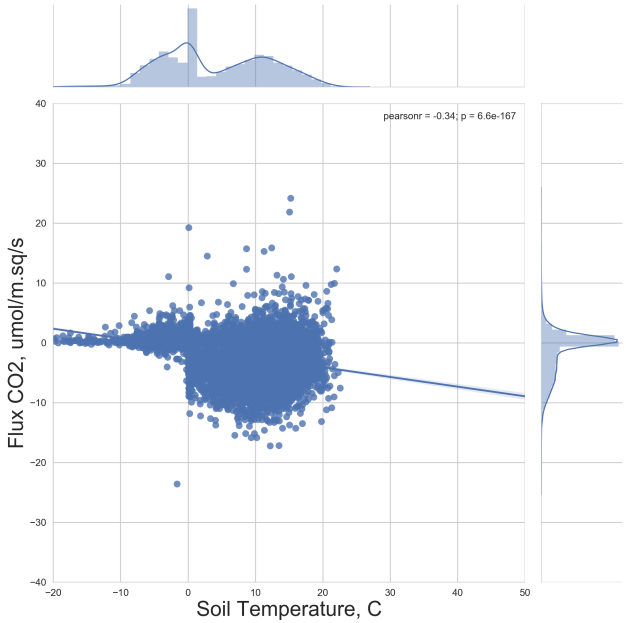


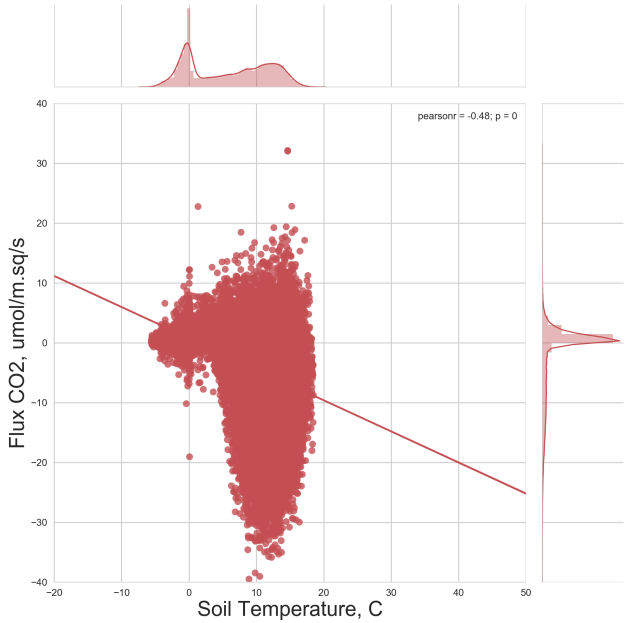
Flux vs Temperature during Daylight

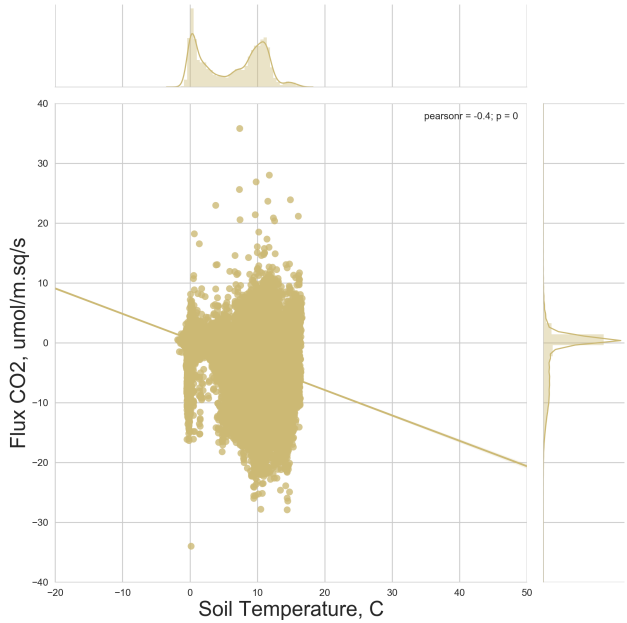




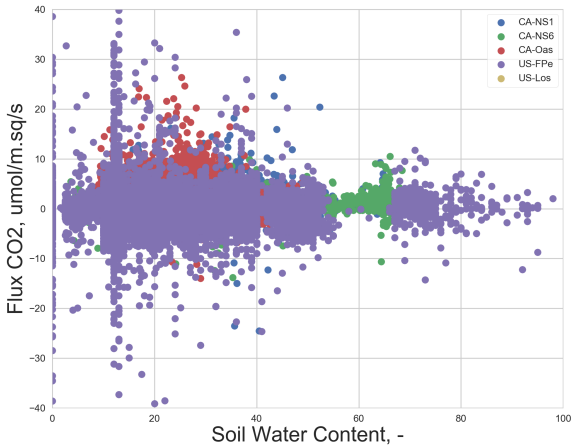


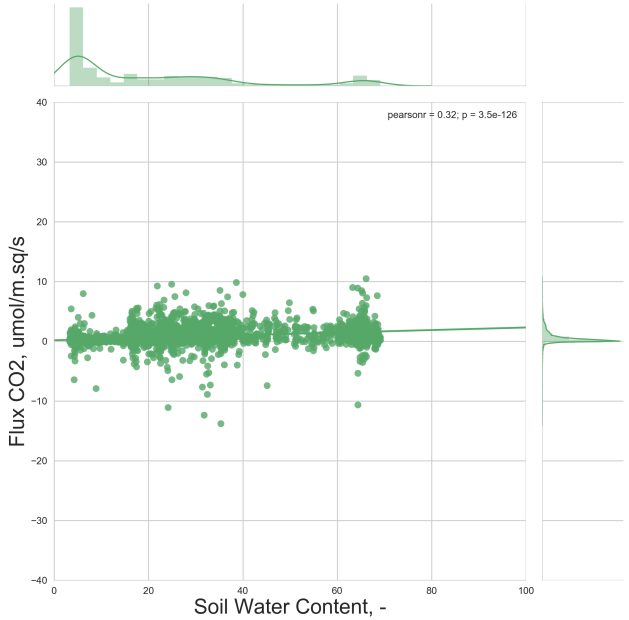


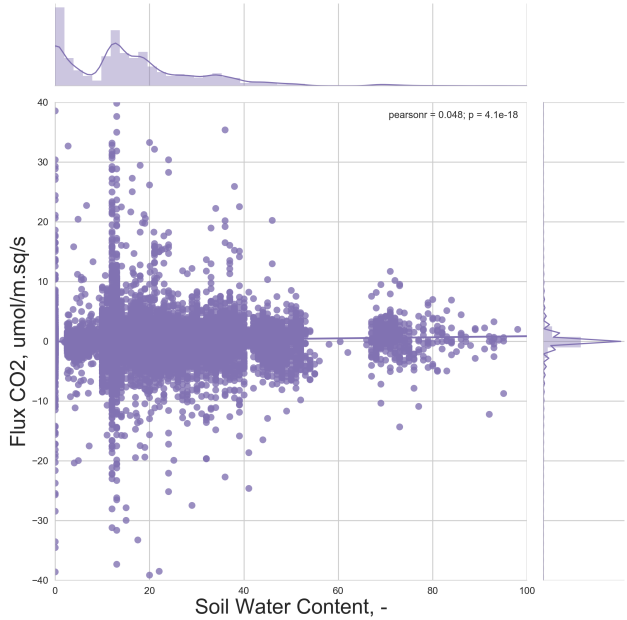


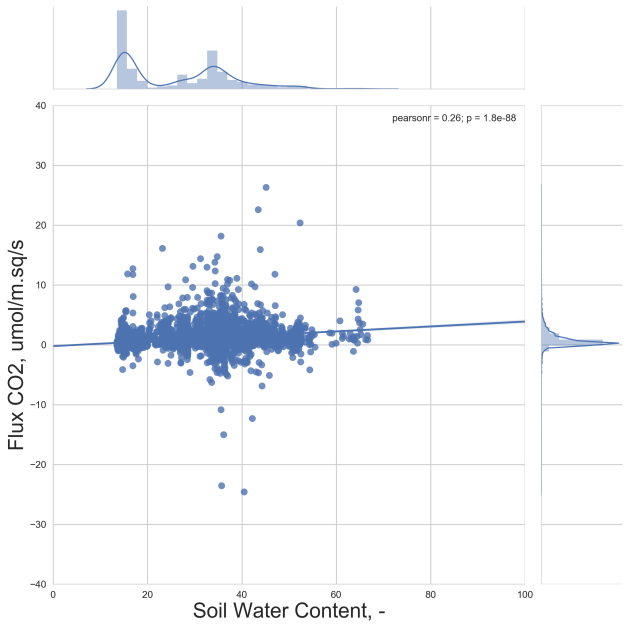


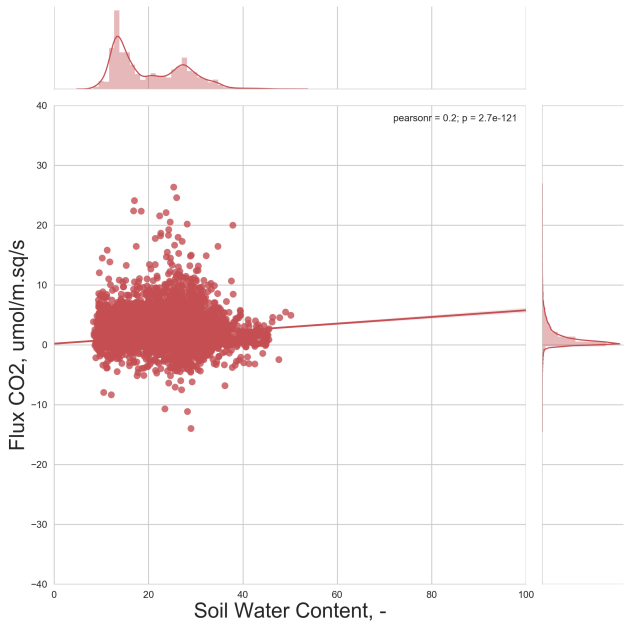
Flux vs Soil Water Content at Night

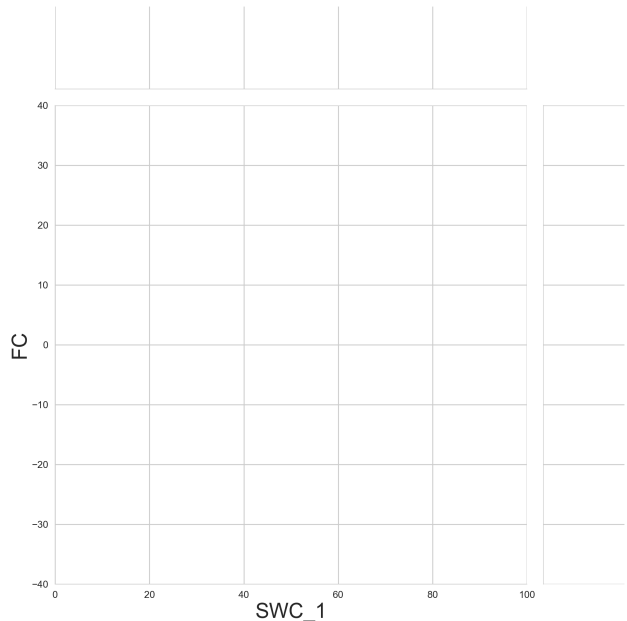




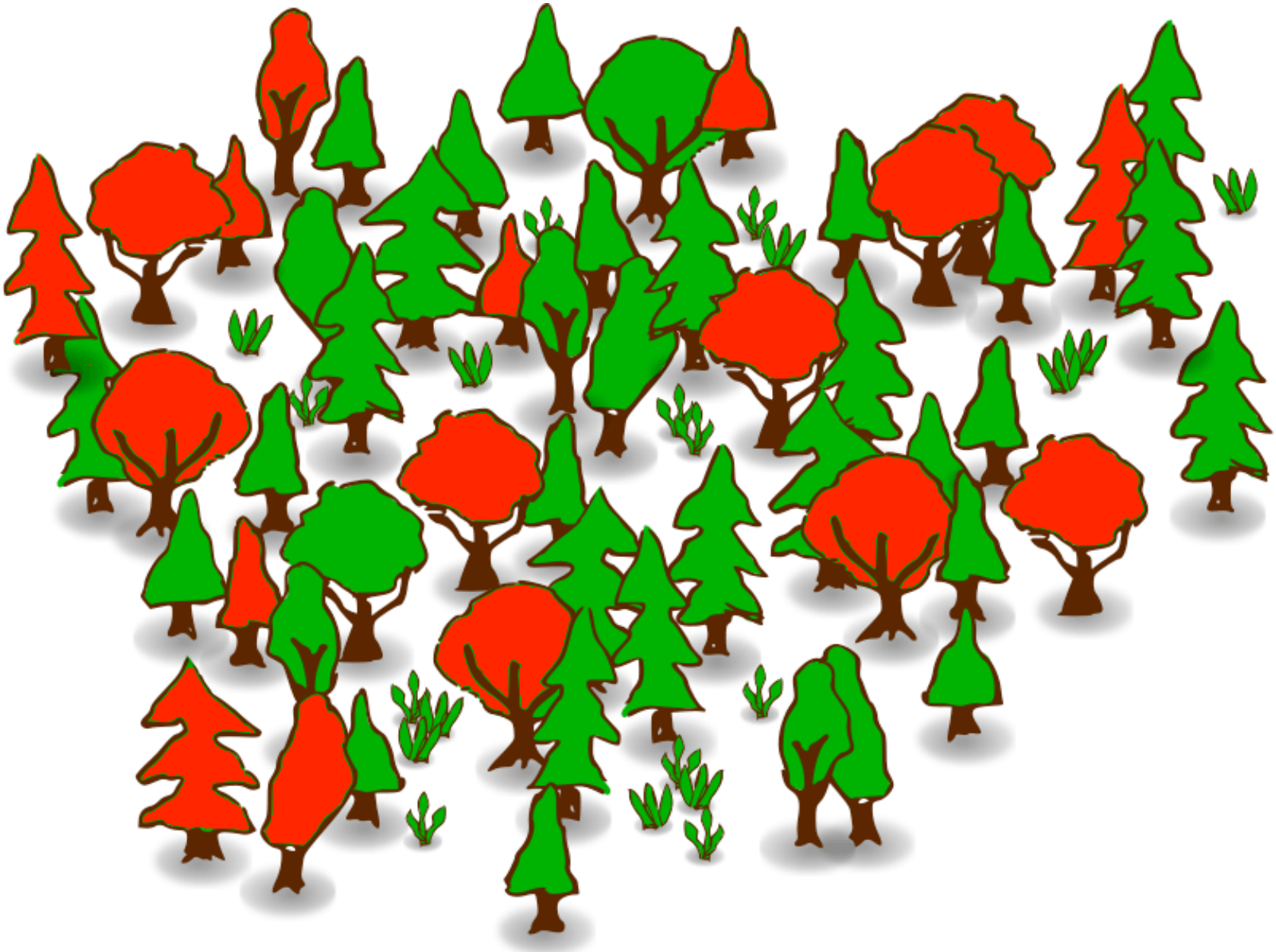






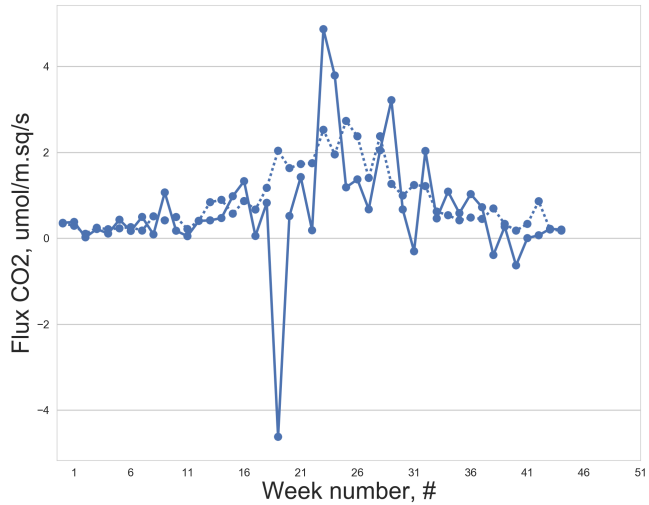


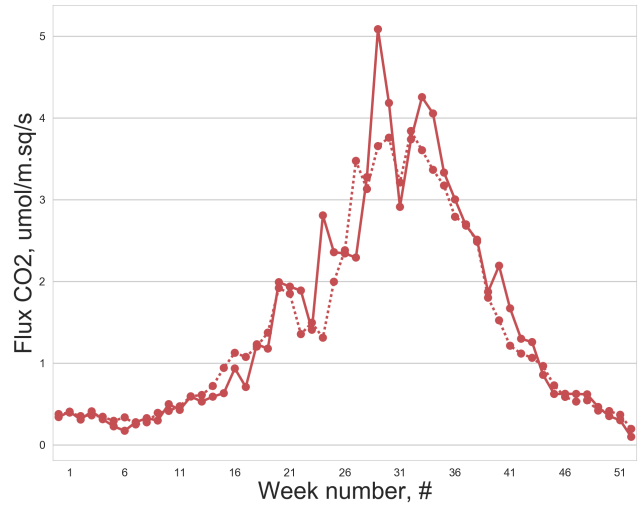
Random Forest Algorithm

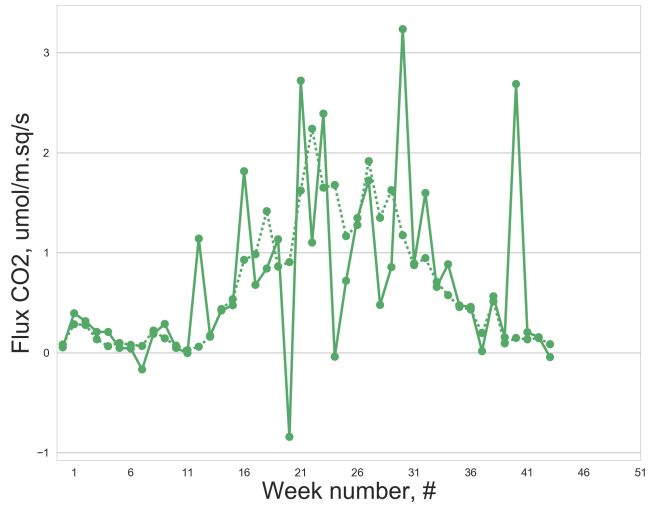


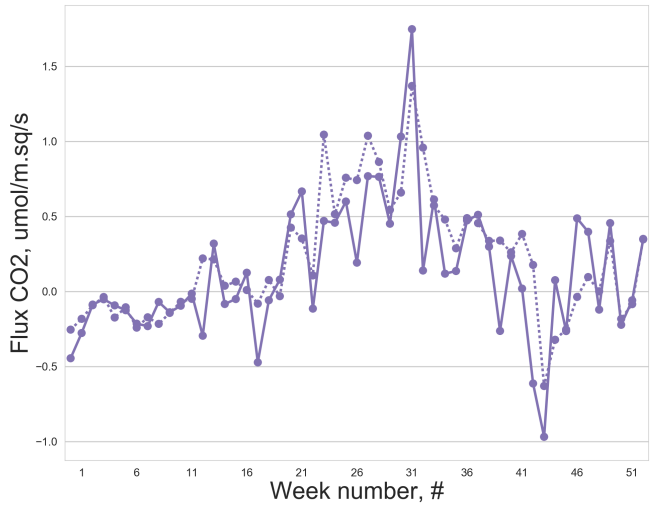
Random Forest Algorithm: Predicted vs Measured at Night

Features: SWC, T, P, Lat, Long, Date



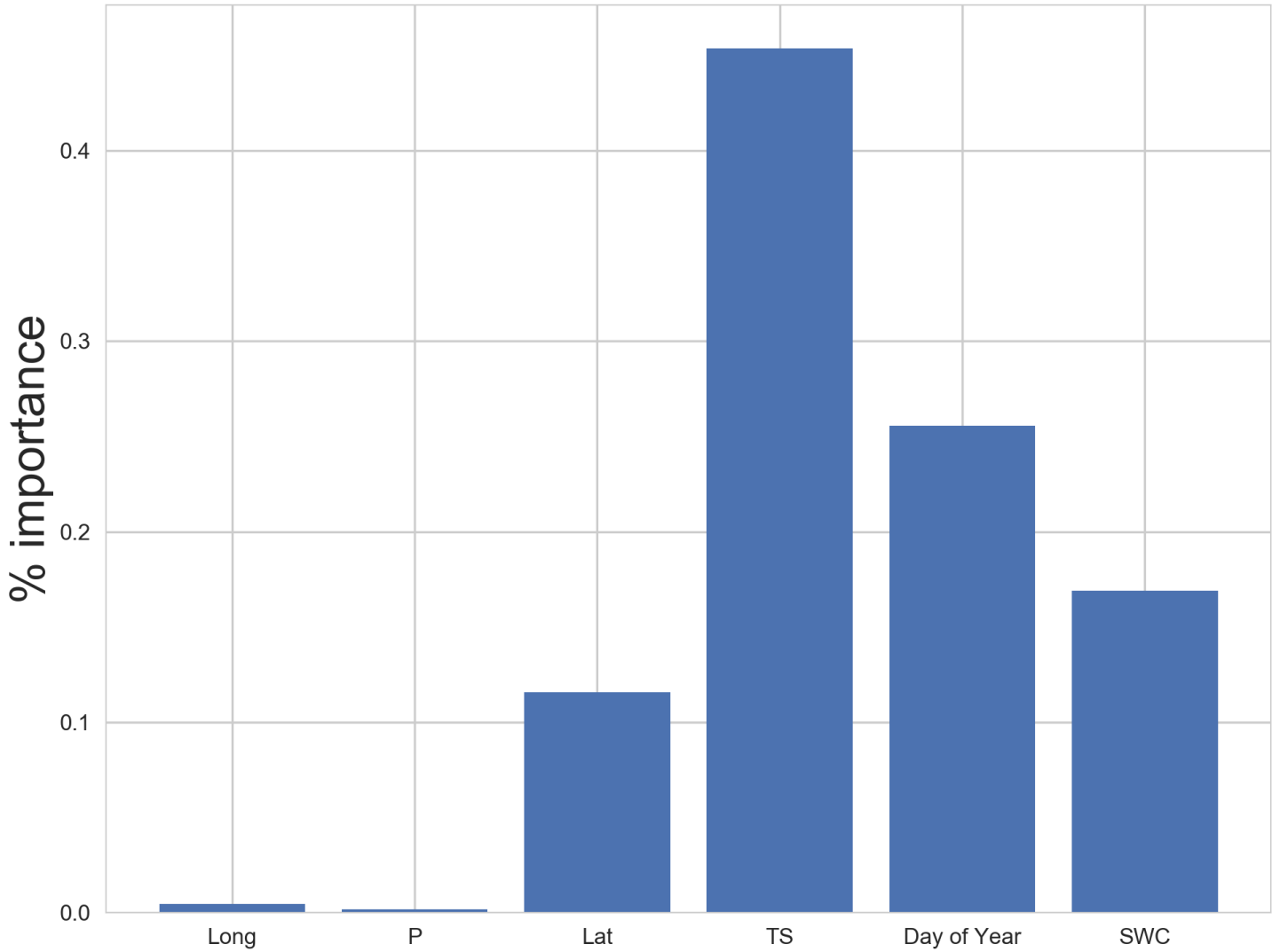






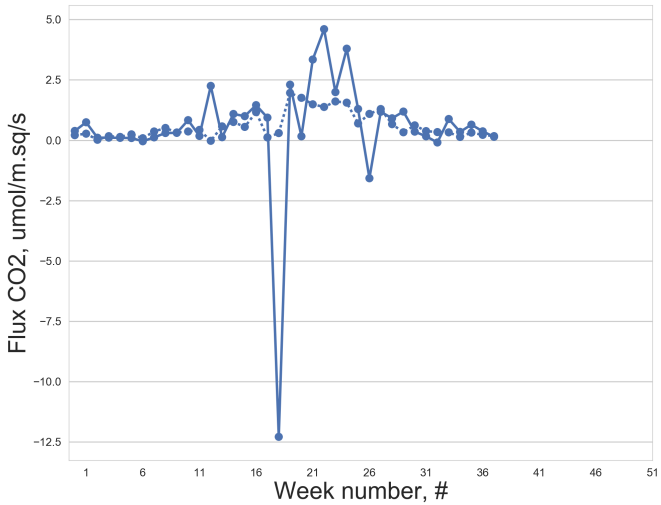
Random Forest Algorithm:

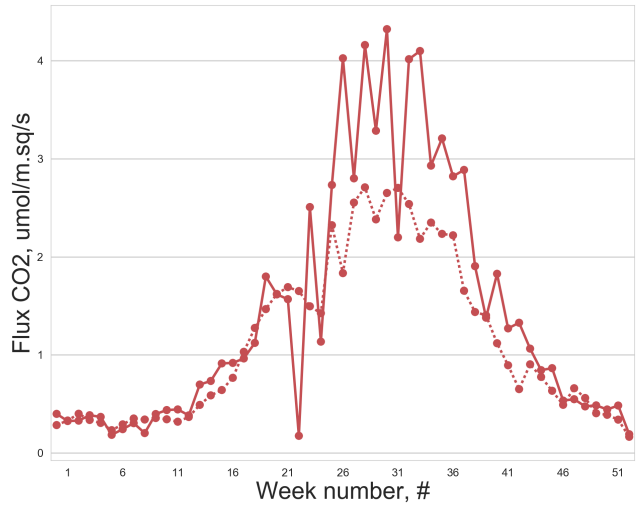
Relative Importance of Features

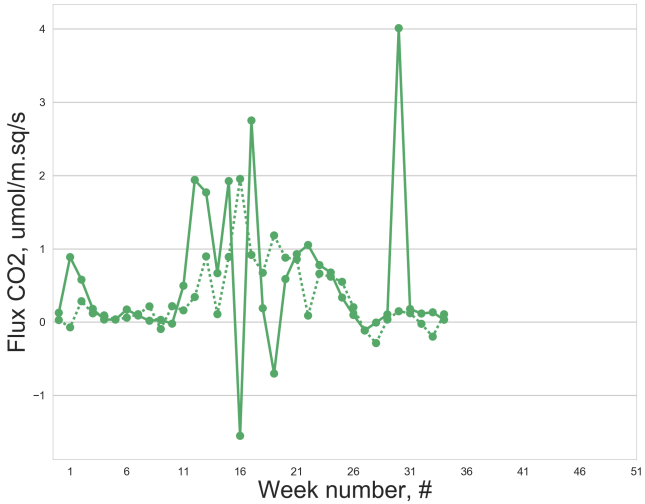


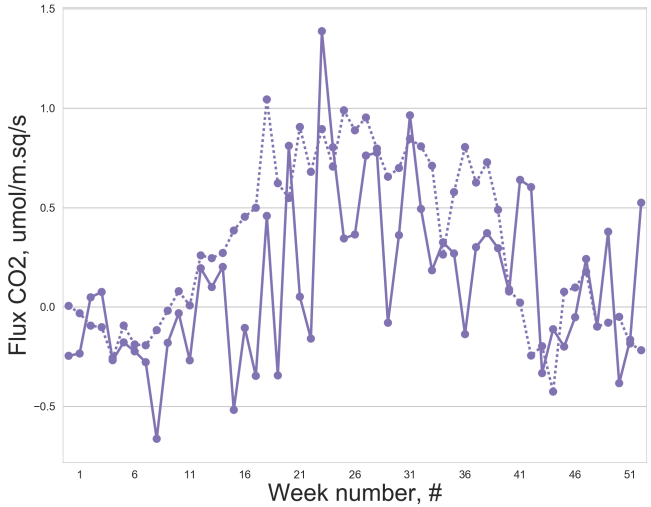
Random Forest Algorithm: Predicted vs Measured at Night

Features: SWC, T



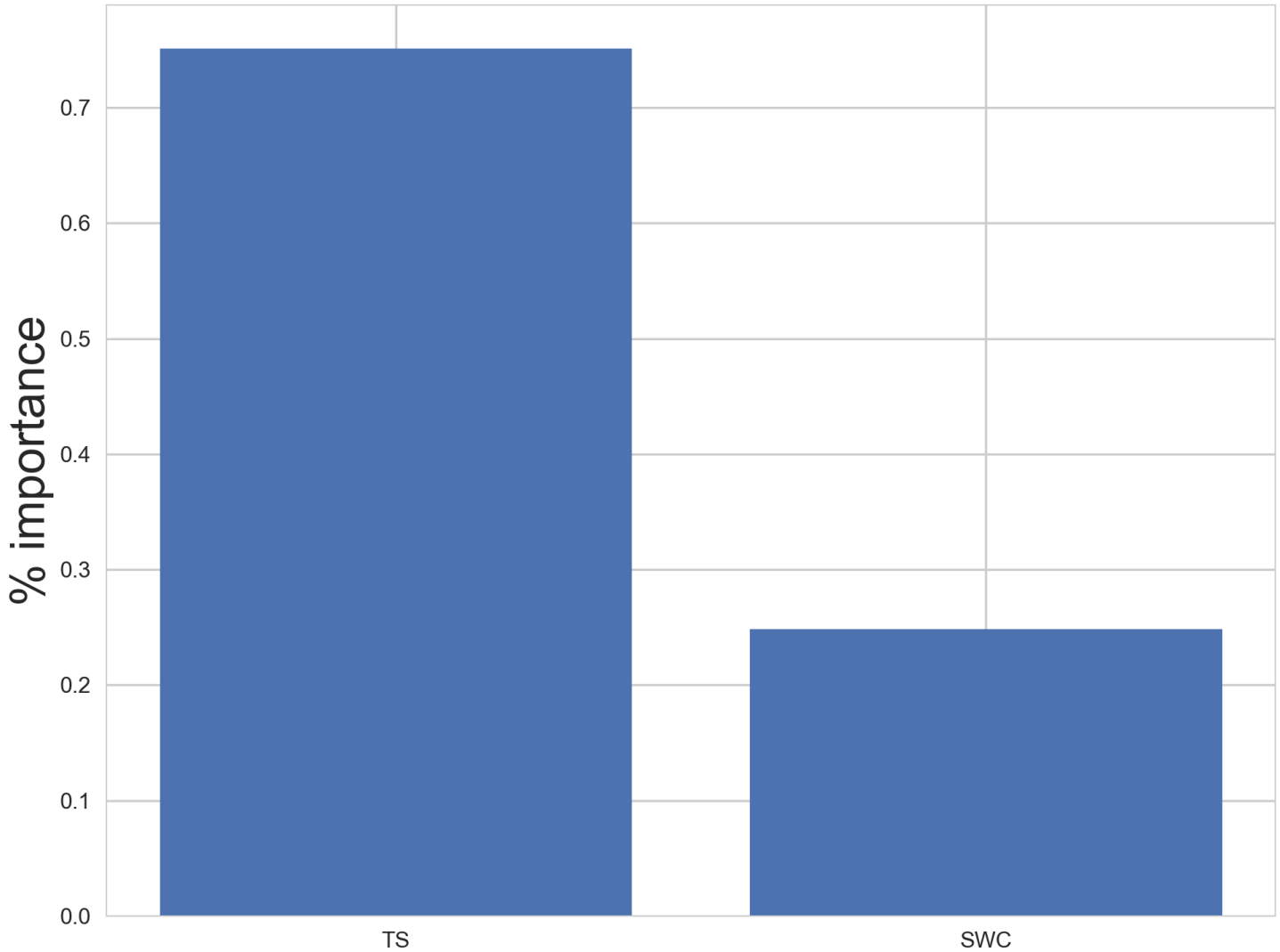






Random Forest Algorithm: Train on all data (25 Sites)

Relative Importance of Features



Train on all data (25 Sites). Six Modeled Sites

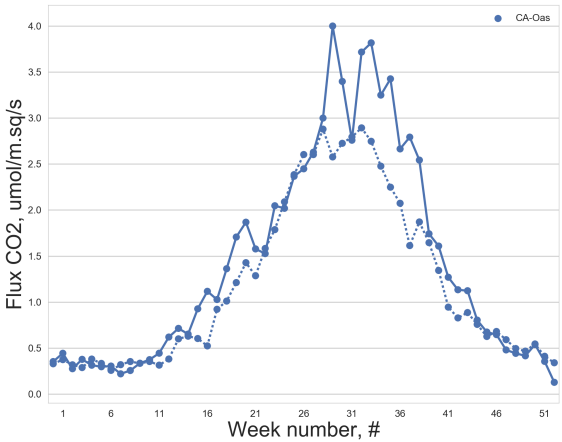
Modeled Sites:

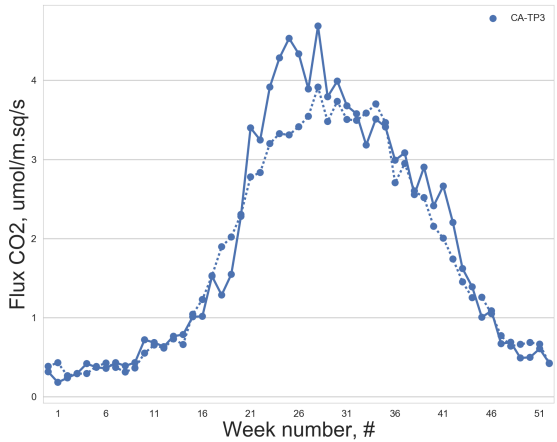
Station	Climate	Type	Mean T, C	Mean P, mm	Elev., m
CA-OAS	Dfc	DBF	0.34	428	530
CA-Qfo	Dfc	ENF	-0.36	962	382
CA-TP3	Dfb	ENF	8	1036	184
US-Fpe	Bsk	GRA	5.48	335	634
CA-TPD	Bsk	GRA	8	1036	260
CA-GRO	Dfb	MF	1.3	831	340

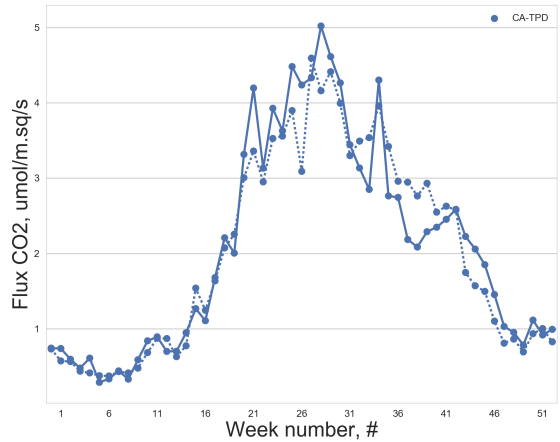
Climate: Dfc - Subarctic: severe winter, no dry season, cool summer; Bsk - Cold semi-arid climate, steppe, warm winter; Dfb - Warm Summer Continental: significant precipitation in all seasons.
Vegetation: ENF - Evergreen Needleleaf Forests; DBF - Deciduous Broadleaf Forests; GRA - grassland; MF - mixed forest;

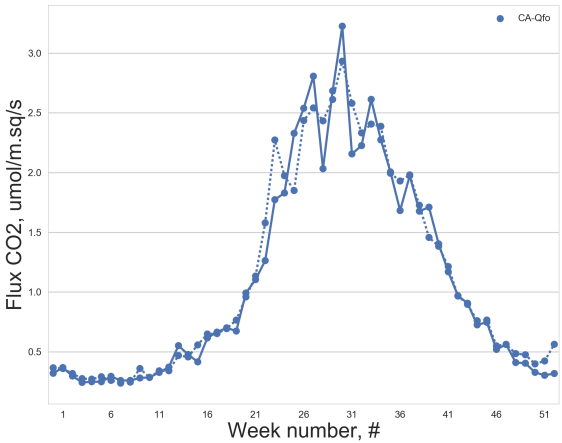
Random Forest Algorithm: Predicted vs Measured at Night

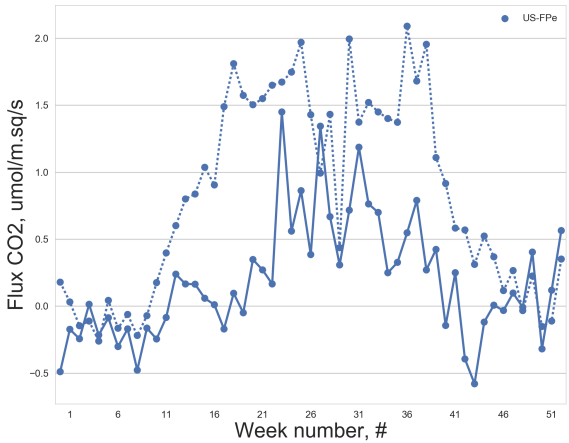
Features: SWC, T; Train on all data

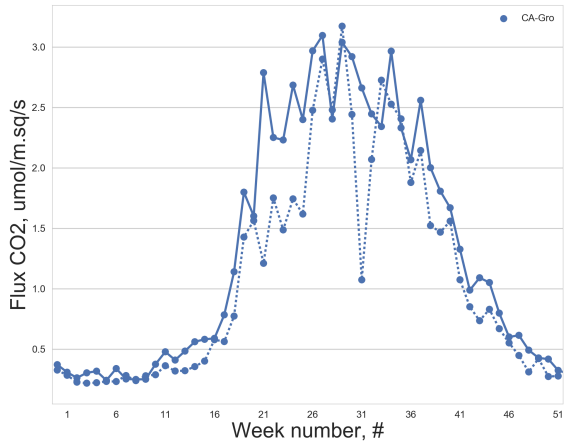






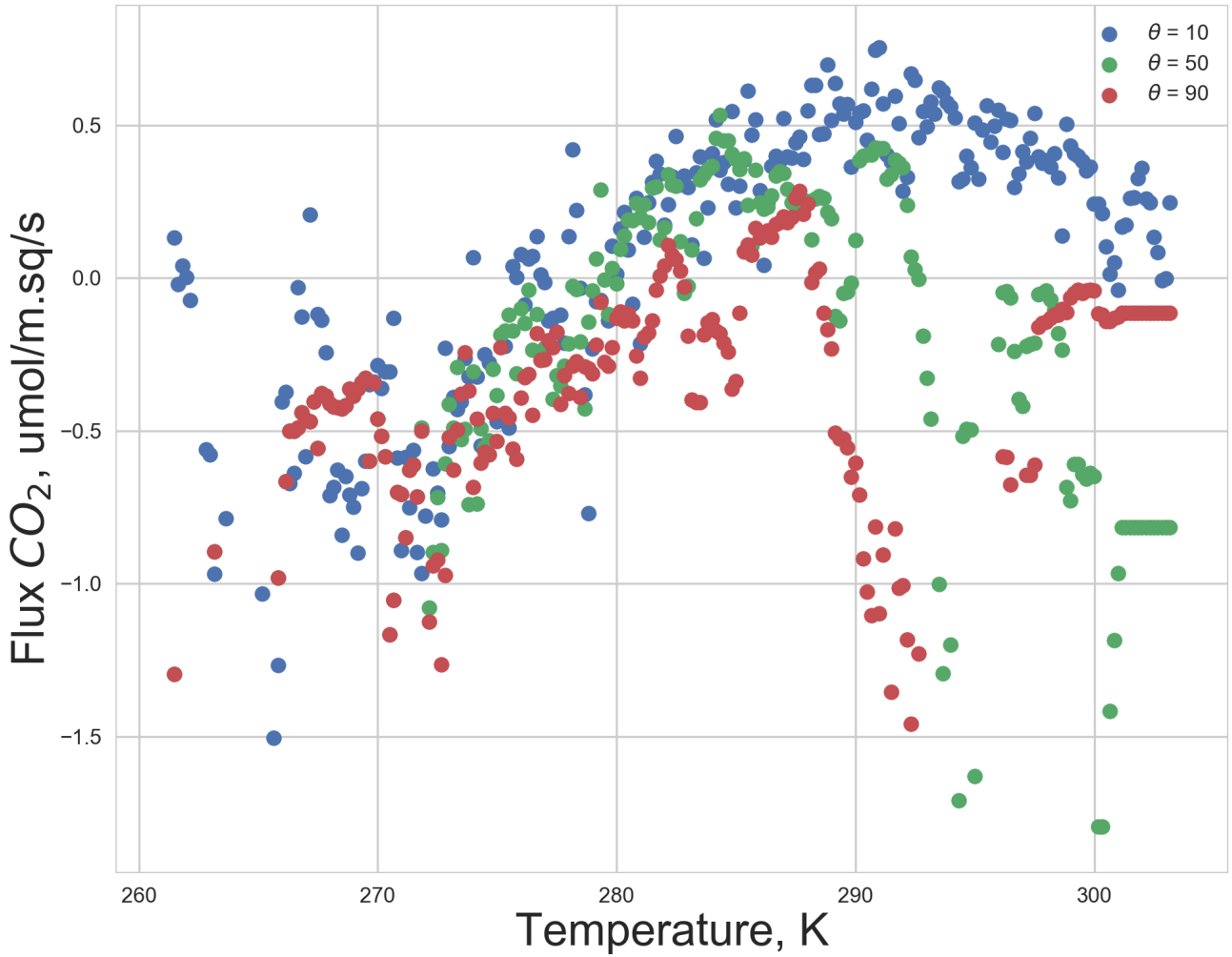






Reverse Engineering of Random Forest Tree Algorithm

T dependence



Reverse Engineering of Random Forest Tree Algorithm

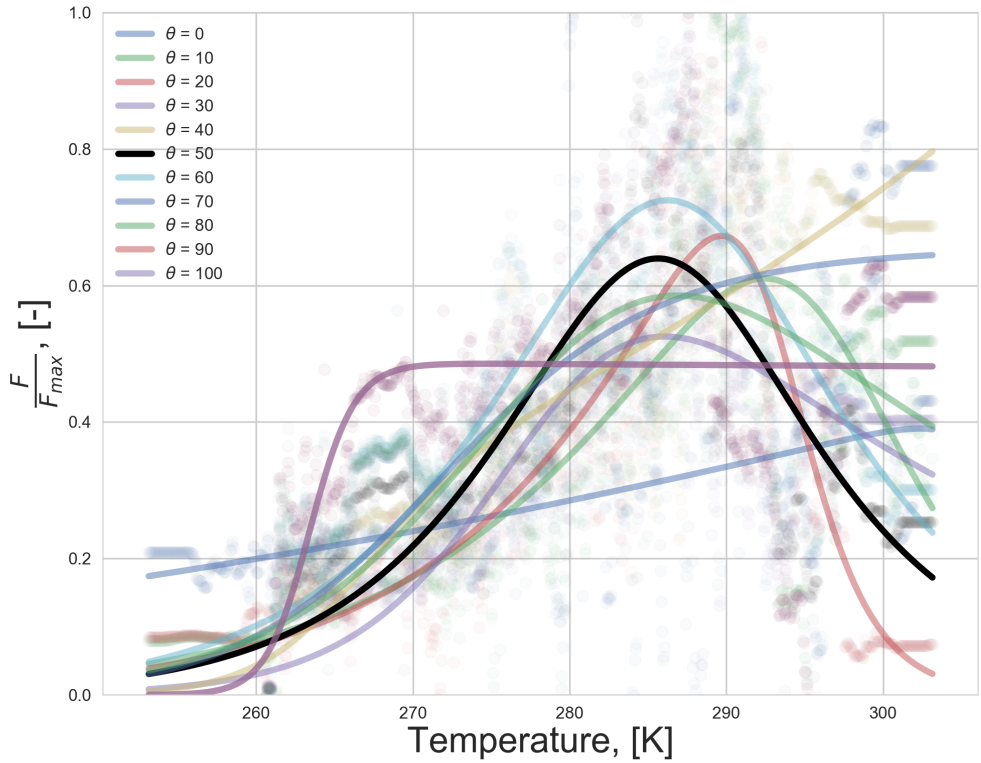
T dependence

Modified Arrhenius equation for optimum T:

$$\frac{F}{F_{max}} = \frac{exp\left[\frac{-E_a}{RT_0}\left(1-\frac{T_0}{T}\right)\right]}{1+exp\left[\frac{ST-H_d}{RT}\right]} \quad (1)$$

Fitting params:

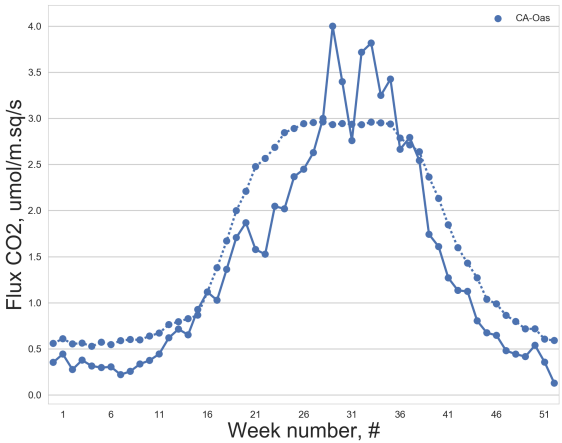
$$T_0 = 284 \text{ K} \quad (2)$$
$$E_a = -67.5 \text{ kJ/mol} \quad (3)$$
$$S = 543.6 \text{ J/mol/K} \quad (4)$$
$$H_d = 155.9 \text{ kJ/mol} \quad (5)$$

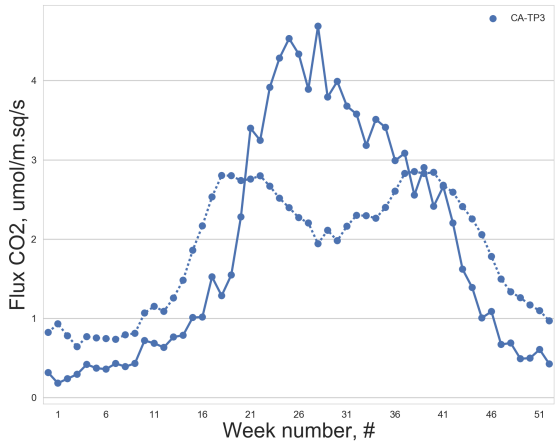


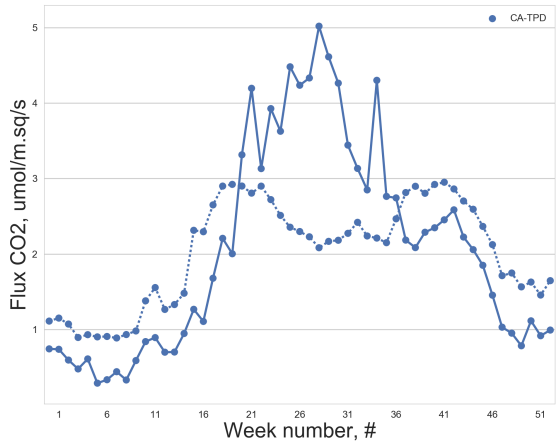
Reverse Engineering: Predicted (eq. 6) vs Measured

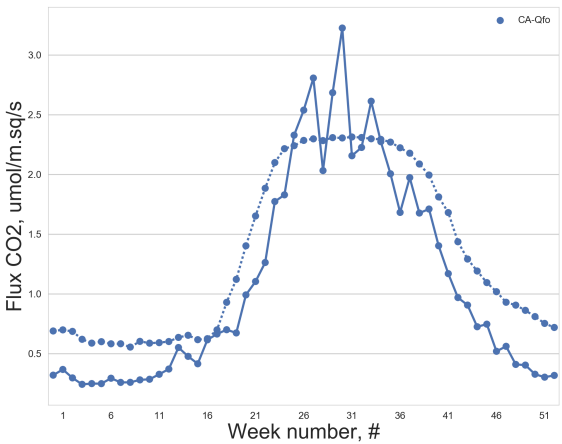
Measured - solid line; eq. 6 - dashed line

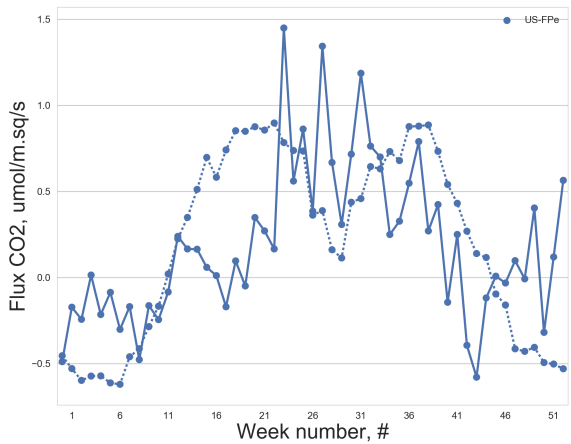
$$F = F_{max} \cdot f(T) \tag{6}$$

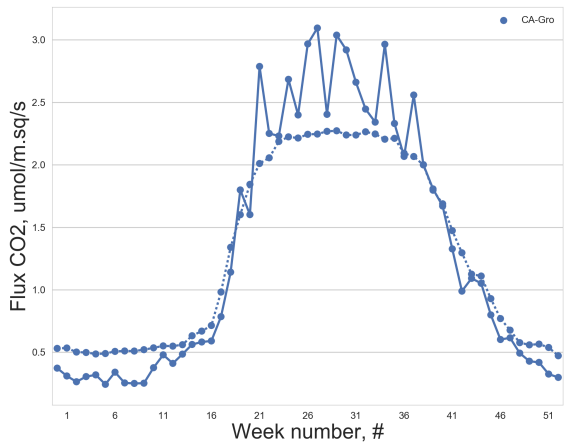






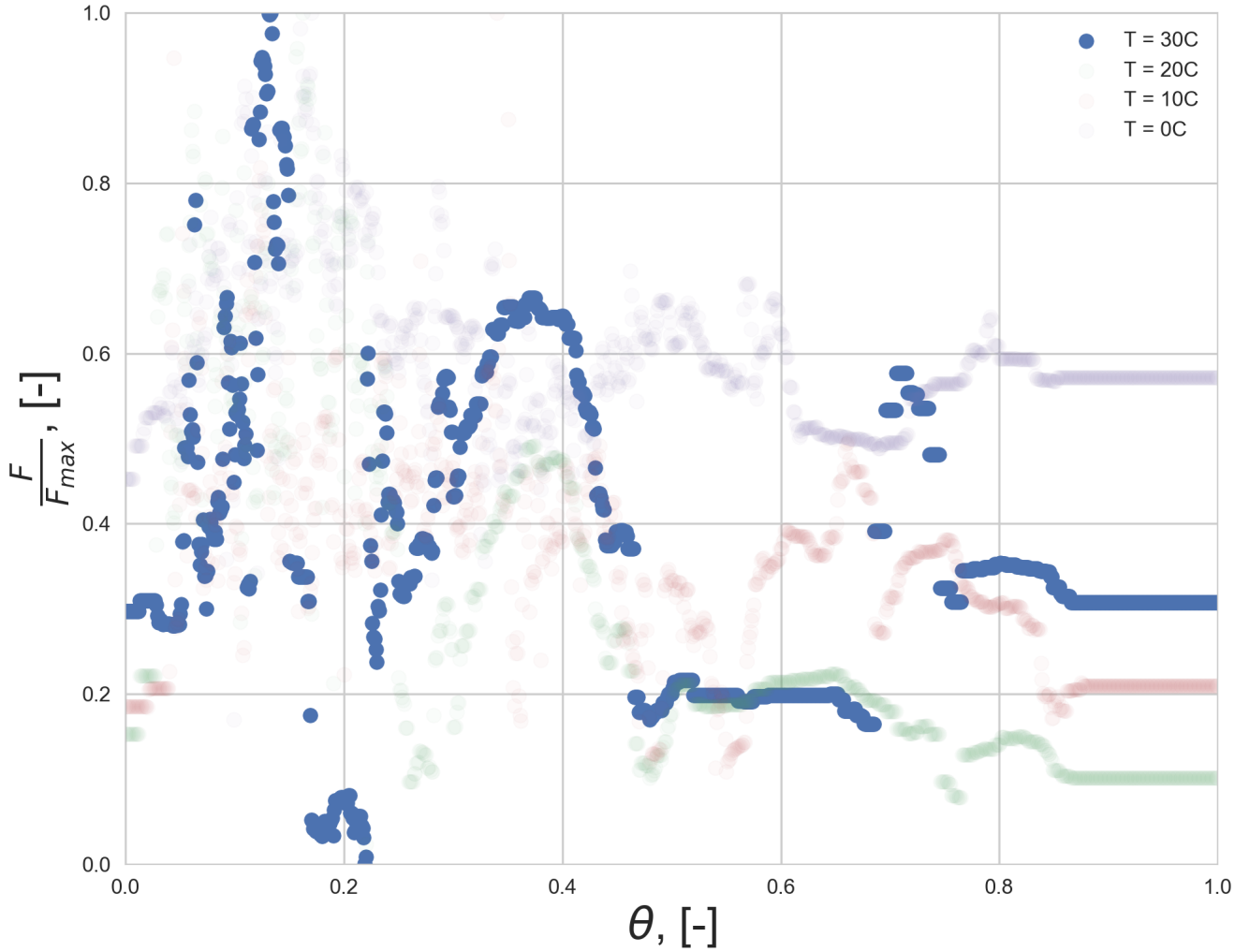






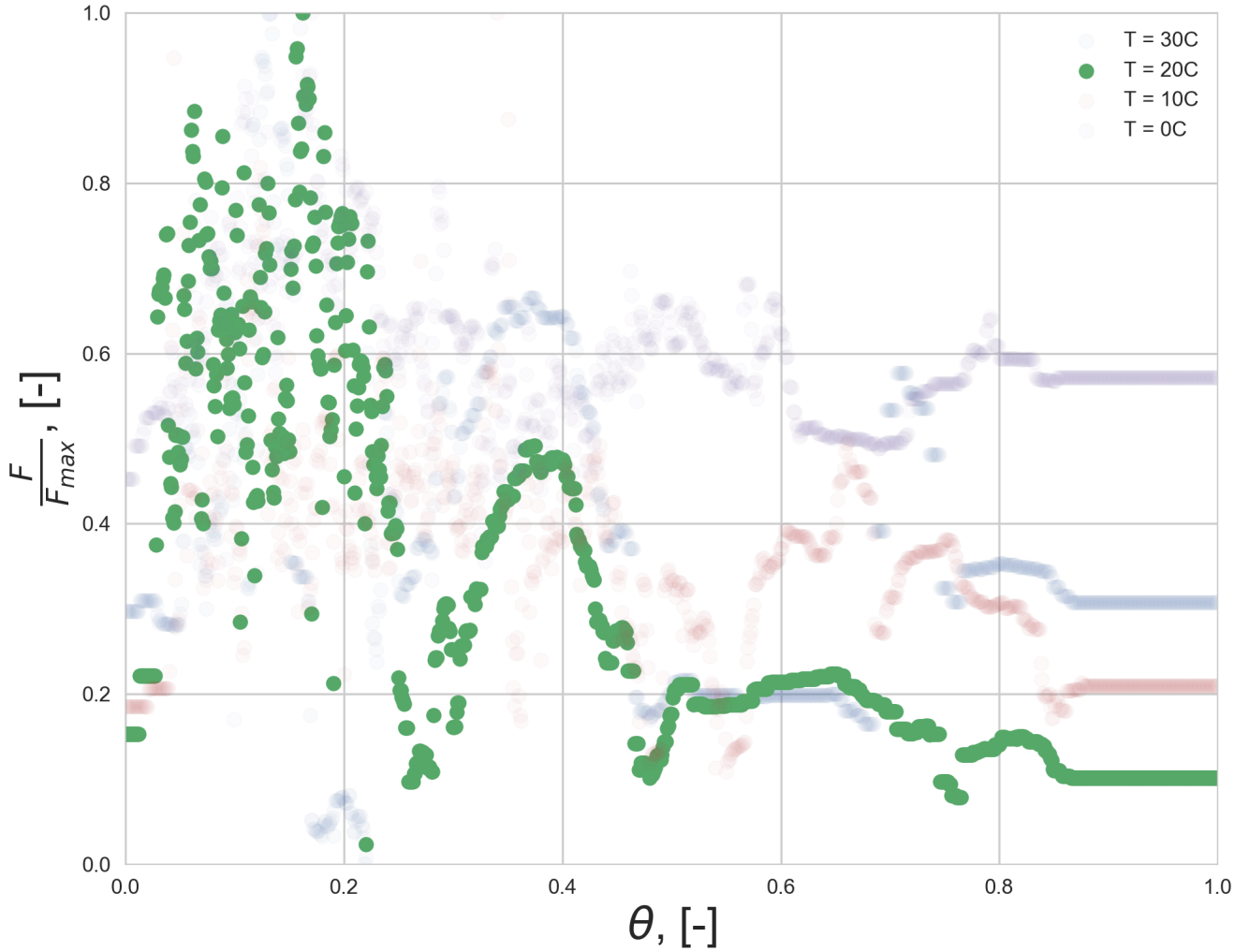
Reverse Engineering of Random Forest Tree Algorithm

θ dependence



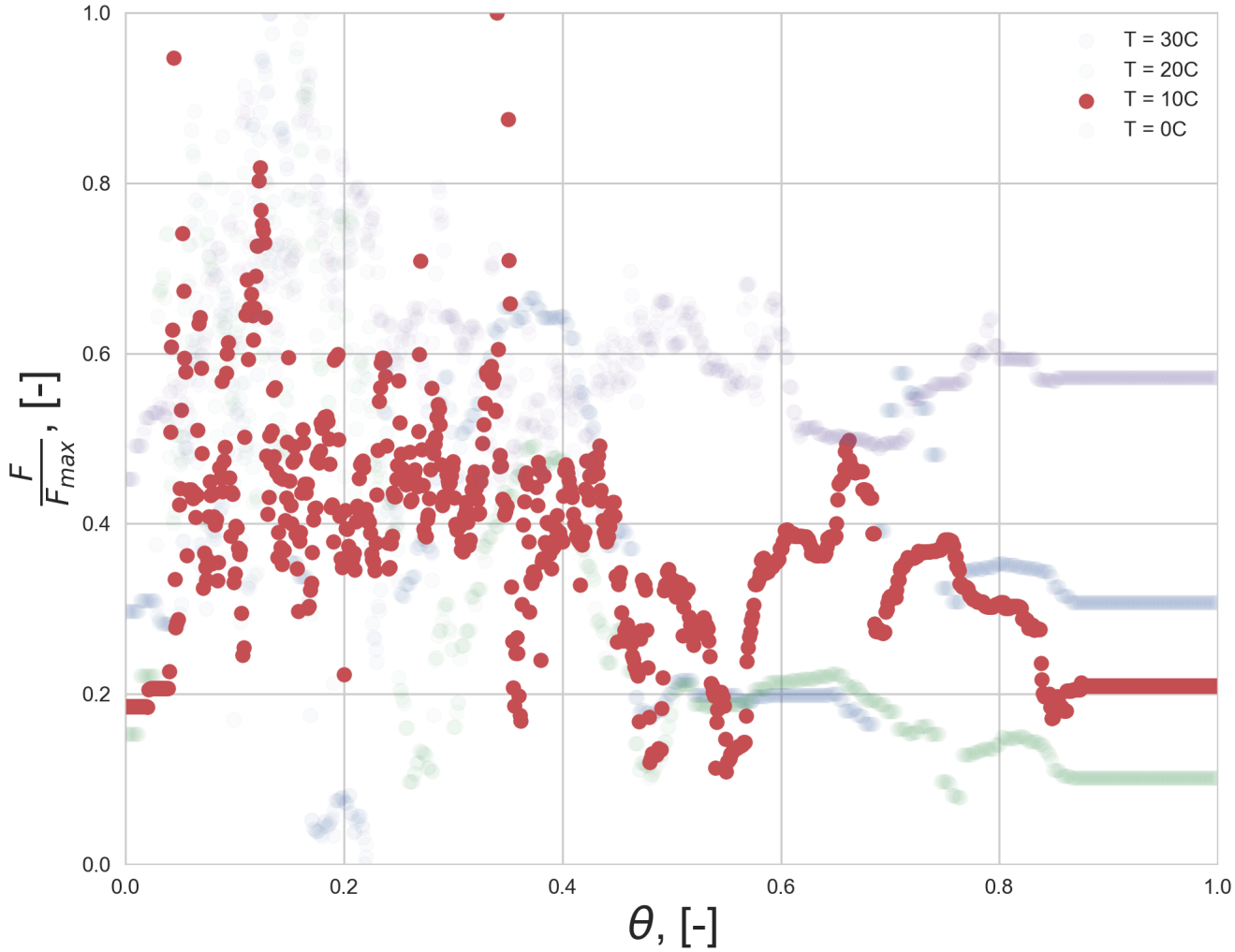
Reverse Engineering of Random Forest Tree Algorithm

θ dependence



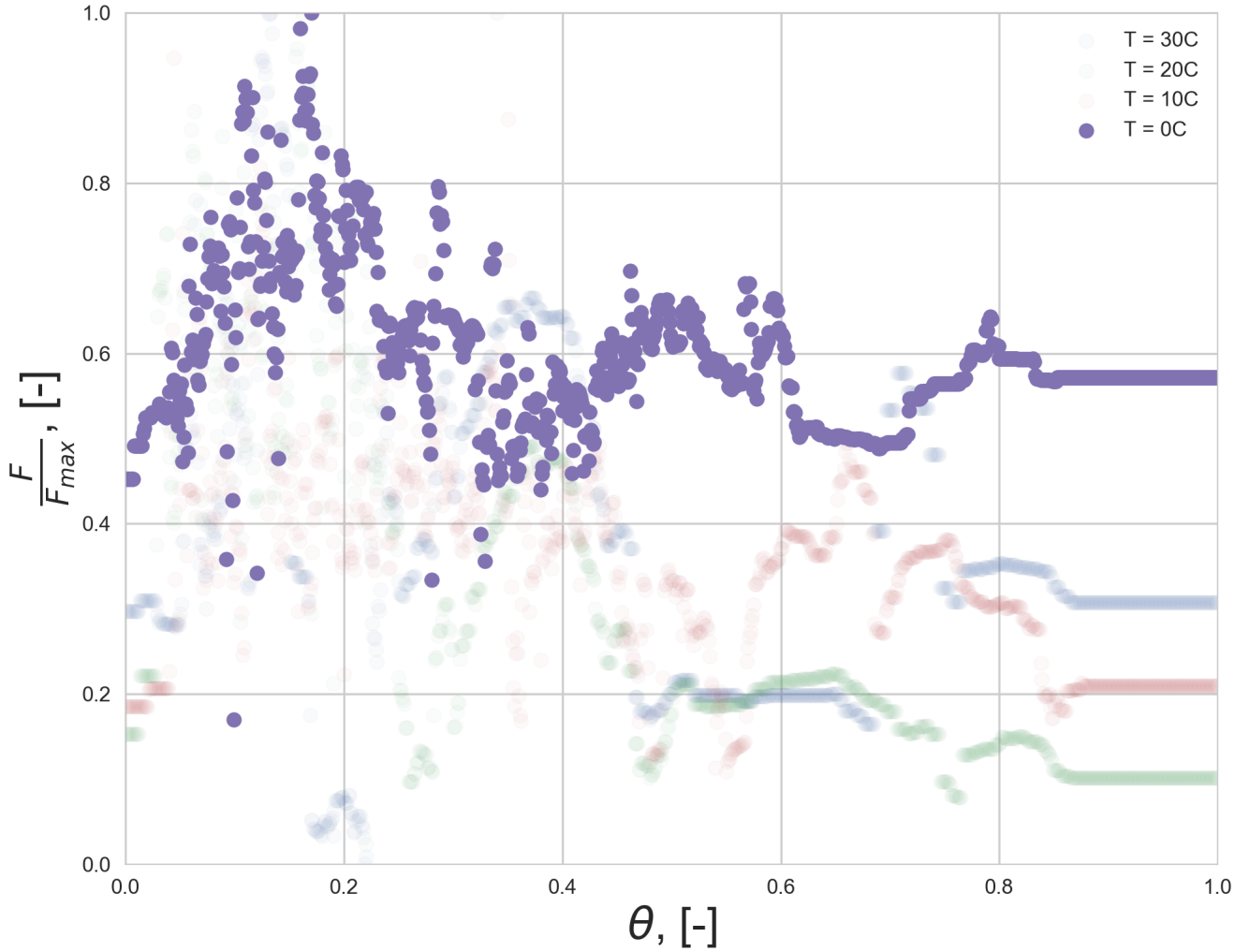
Reverse Engineering of Random Forest Tree Algorithm

θ dependence



Reverse Engineering of Random Forest Tree Algorithm

θ dependence



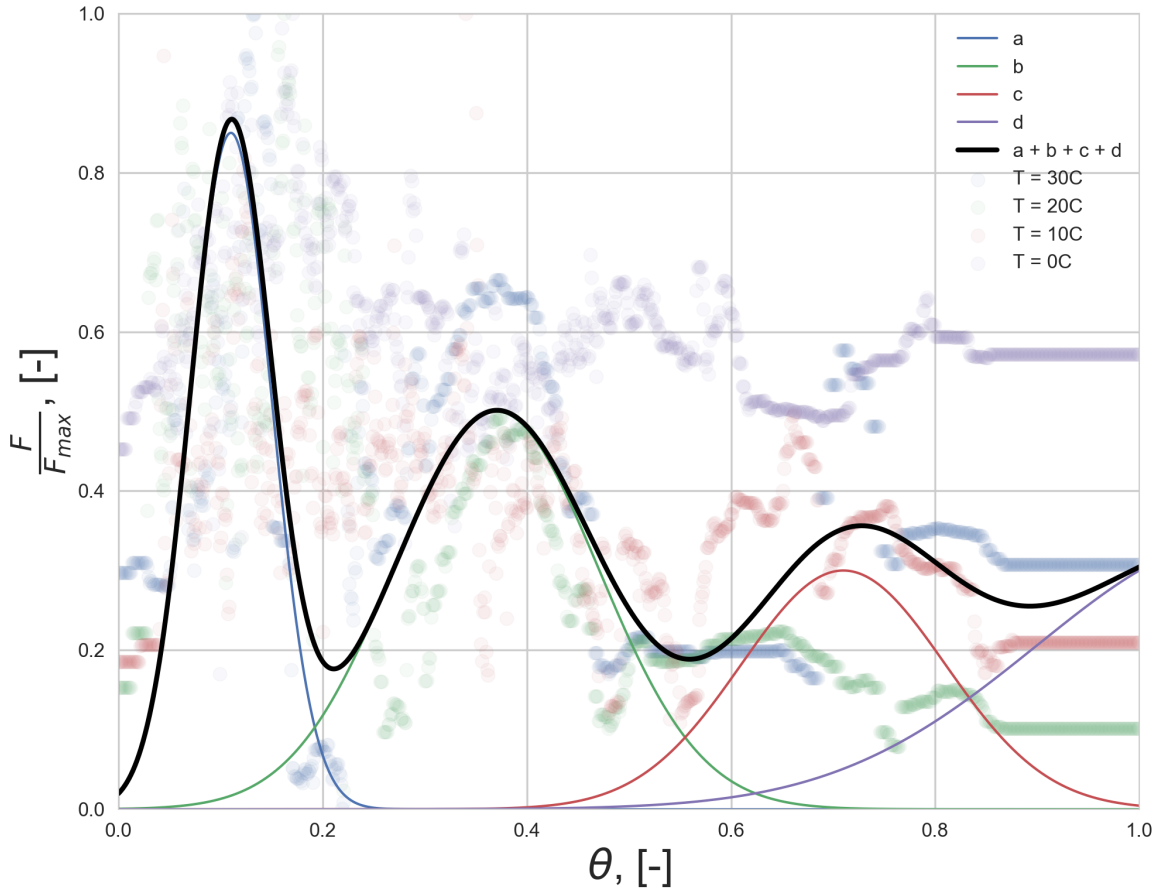
Reverse Engineering of Random Forest Tree Algorithm

θ dependence

$$\frac{F}{F_{max}} = \sum_{i}^{4} A_i \frac{1}{\sqrt{2\pi\sigma_i^2}} e^{-\frac{(x-\mu_i)^2}{2\sigma_i^2}} \qquad (7)$$

Fitting params:

Parameter	a	b	c	d
A	0.85	0.5	0.5	0.3
μ	0.11	0.37	0.71	1.1
σ	0.04	0.1	0.1	0.2



Reverse Engineering: Predicted (eq. 8) vs Measured

Measured - solid line; eq. 8 - dashed line

$$F = F_{max} \cdot f(T) \cdot f(\theta) \tag{8}$$