Identification and Prediction of Flux Tower Latent Heat Data and Their Source Variables (Time Series Imputation)

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Dataset Description

We propose to use Goose Creek Eddy Covariance Flux Tower Sensor Data[kumar2024?]. The data is collected from the Eddy Covariance Flux Tower in Goose Creek, Piatt County. The dataset consists of time series data spanning from Spring 2016 to Spring 2023 with 15 minutes time interval. Dataset involves 167 variables shown in Table 1 including latent heat, sensible heat, wind speed, temperature, and changes in the ecosystem with respect to water, carbon, and temperature. Figure 1 illustrates part of variables in 2022. The data collected by flux tower provides a foundation for further investigation into hydrological, meteorological, and environmental phenomena. The format of dataset is CSV file (generated from raw PICKLE file). The dataset can be found through link:

https://www.hydroshare.org/resource/c276c71e8d1246e29d8502f5b2054668/

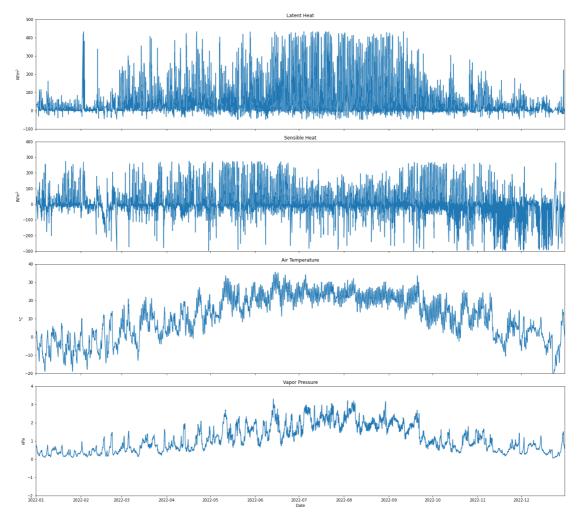


Fig 1: Variables Latent Heat, Sensible Heat, Air Pressure and Vapor Pressure in 2022

Table 1: A table with a variables.

Variable Name	Units	Description	
TIMESTAMP	TS		
RECORD	RN		
Hs	W/m^2	sensible heat flux	
tau	$kg/(m \cdot s^2)$	shear stress	

Variable Name	Units	Description	
u_star	m/s	friction velocity	
Ts_stdev	$^{\circ}C$	instantaneous stdev of temperature	
Ts_Ux_cov	$^{\circ}C\cdot m/s$	inst. cov(temp, Ux)	
Ts_Uy_cov	$^{\circ}C\cdot m/s$	inst. cov(temp, Uy)	
Ts_Uz_cov	$^{\circ}C\cdot m/s$	inst. cov(temp, Uv)	
Ux_stdev	m/s	instantaneous stdev of Ux	
Ux_Uy_cov	$(m/s)^2$	instantaneous cov of (Ux, Uy)	
Ux_Uz_cov	$(m/s)^2$	instantaneous cov of (Ux, Uz)	
Uy_stdev	m/s	instantaneous stdev of Uy	
Uy_Uz_cov	$(m/s)^2$	instantaneous cov of (Uy, Uz)	
Uz_stdev	m/s	instantaneous stdev of Uz	
wnd_spd	m/s	wind speed (horizontal) - different from next?	
rslt_wnd_spd	m/s	wind speed (horizontal)	
wnd_dir_sonic	degrees	wind direction from CSAT3, deg from N?	
std_wnd_dir	degrees	inst. stdev of wind direction	
wnd_dir_compa ss	degrees	wind direction from compass (from N?)	
Ux_Avg	m/s	average horiz windspeed x	
Uy_Avg	m/s	average horiz windspeed y	
Uz_Avg	m/s	average vertical windspeed z	
Ts_Avg	$^{\circ}C$	air temperature at 25 m	
sonic_azimuth	degrees	180 is direction is pointing - can change this value	
sonic_samples_ Tot	samples	10 Hz sampling rate (cycles per 15 mins = 9000)	
Fc_li_wpl	$mg/(m^2 \ \cdot s)$	carbon flux upward (+ = upward) with Webb et al Term	
LE_li_wpl	W/m^2	latent heat flux with Webb et al term	
Hc_li	W/m^2	sensible heat flux	
CO2_li_mean	mg/m^3	CO2 conc -> need to convert to ppm units	
H2O_li_mean	g/m^3	water vapor conc at 25 m	
amb_press_li_m ean	kPa	air pressure at 25 m	
Tc_li_mean	$^{\circ}C$	CSAT air temperature at 25 m	
rho_a_li_mean	kg/m^3	density of air with water vapor	
Fc_li_irga	$mg/(m^2 \ \cdot s)$	carbon flux without Webb et al. Term	

Variable Name	Units	Description	
LE_li_irga	W/m^2	latent heat flux without Webb et al. Term	
irga_li_samples_ Tot	samples	should be around 60 - quality indicator of LiCor	
Precip_Tot	mm	rainfall	
T_tmpr_rh_mea n	$^{\circ}C$	air temperature at 25 m	
e_tmpr_rh_mea n	kPa	vapor pressure at 25 m	
e_sat_tmpr_rh_ mean	kPa	saturated vapor pressure at 25 m	
H2O_tmpr_rh_ mean	g/m^3	water vapor conc at 25 m	
RH_tmpr_rh_me an		Relative Humidity at 25 m (e/e_sat)	
rho_a_tmpr_rh_ mean	kg/m^3	air density	
slowsequence_ 1_Tot	samples	cycles per 15 mins - scanning every 10 secs	
CS655_Wcr_Avg	m^3/m^3	soil water content	
CS655_Ec_Avg	dS/m	soil conductivity	
CS655_Tmpr_Av	$^{\circ}C$	soil temperature	
mean_wind_spe ed	m/s	wind speed at 10 m heightnot average?	
mean_wind_dir ection	degrees	wind direction	
std_wind_dir	degrees	mean wind vector stdev of direction	
NDVI_Avg		Normalized Difference Vegetation Index	
NDVIUpRed_Av g	$W/m^2 \ \cdot nm$	NDVI is calculated from upward and canopy facing sensors that measure IR and NIR radiation	
NDVIUpNIR_Avg	$W/m^2 \ \cdot nm$	NDVI is calculated from upward and canopy facing sensors that measure IR and NIR radiation	
NDVIIndUp		NDVI is calculated from upward and canopy facing sensors that measure IR and NIR radiation	
NDVIDownRed_ Avg	$W/m^2 \over \cdot nm$	NDVI is calculated from upward and canopy facing sensors that measure IR and NIR radiation	
NDVIDownNIR_ Avg	$W/m^2 \over \cdot nm$	NDVI is calculated from upward and canopy facing sensors that measure IR and NIR radiation	
NDVIIndDown		NDVI is calculated from upward and canopy facing sensors that measure IR and NIR radiation	
PRI_Avg		Photochemical Reflectance Index	
PRIUp531_Avg	$W/m^2 \over \cdot nm$	PRI calculated from updward and canopy facing sensors that measure 2 wavelengths of radiation	

Variable Name	Units	Description	
PRIUp570_Avg	$W/m^2 \ \cdot nm$	PRI calculated from updward and canopy facing sensors that measure 2 wavelengths of radiation	
PRIIndUp		PRI calculated from updward and canopy facing sensors that measure 2 wavelengths of radiation	
PRIDown531_Av g	$W/m^2 \ \cdot nm$	PRI calculated from updward and canopy facing sensors that measure 2 wavelengths of radiation	
PRIDown570_Av	$W/m^2 \ \cdot nm$	PRI calculated from updward and canopy facing sensors that measure 2 wavelengths of radiation	
PRIIndDown		PRI calculated from updward and canopy facing sensors that measure 2 wavelengths of radiation	
D5TE_VWC_5cm _Avg	m^3/m^3	volumetric water content	
D5TE_P_5cm_Av g		bulk dielectric permittivity	
D5TE_EC_5cm_A vg	dS/m	soil electrical conductivity	
D5TE_T_5cm_Av	$ ceil^{\circ}C$	soil temperature	
D5TE_VWC_15c m_Avg	m^3/m^3	volumetric water content	
D5TE_P_15cm_A vg		bulk dielectric permittivity	
D5TE_EC_15cm_ Avg	dS/m	soil conductivity	
D5TE_T_15cm_A vg	$ centricle{}^{\circ}C$	soil temperature	
D5TE_VWC_30c m_Avg	m^3/m^3	volumetric water content	
D5TE_P_30cm_A vg		bulk dielectric permittivity	
D5TE_EC_30cm_ Avg	dS/m	soil conductivity	
D5TE_T_30cm_A vg	$igc ^{\circ}C$	soil temperature	
D5TE_VWC_50c m_Avg	m^3/m^3	volumetric water content	
D5TE_P_50cm_A vg		bulk dielectric permittivity	
D5TE_EC_50cm_ Avg	dS/m	soil conductivity	
D5TE_T_50cm_A vg	$ centricle{}^{\circ}C$	soil temperature	
D5TE_VWC_100 cm_Avg	m^3/m^3	volumetric water content	
D5TE_P_100cm_ Avg		bulk dielectric permittivity	

Variable Name	Units	Description	
D5TE_EC_100c m_Avg	dS/m	soil conductivity	
D5TE_T_100cm_ Avg	$^{\circ}C$	soil temperature	
D5TE_VWC_200 cm_Avg	m^3/m^3	volumetric water content	
D5TE_P_200cm_ Avg		bulk dielectric permittivity	
D5TE_EC_200c m_Avg	dS/m	soil conductivity	
D5TE_T_200cm_ Avg	$^{\circ}C$	soil temperature	
slowsequence_ 2_Tot	samples	cycles - 1 minute loops (number of times scanned)	
SB121TempC_A vg	$^{\circ}C$	SB = sensor body, temp of body of sensor	
Targ121TempC_ Avg	$^{\circ}C$	surface temperature	
Targ121mV_Avg	$ $ $^{\circ}C$		
SB1H1TempC_A vg	$ cent{\circ}C$	SB = sensor body, temp of body of sensor	
Targ1H1TempC _Avg	$^{\circ}C$	surface temperature	
Targ1H1mV_Av	$^{\circ}C$		
short_up_Avg	W/m^2	Incoming shortwave radiation detected by the upward facing instrument	
short_dn_Avg	W/m^2	Outgoing shortwave radiation detected by the downward facing instrument	
long_up_Avg	W/m^2	incoming longwave radiation detected by upward facing instrument	
long_dn_Avg	W/m^2	outgoing longwave radiation detected by downward facing instrument	
cnr4_T_C_Avg	$^{\circ}C$	temperature of sensor	
cnr4_T_K_Avg	K	temperature of sensor in Kelvin	
long_up_corr_A vg	W/m^2	Incoming longwave radiation detected by the upward facing instrument, corrected	
long_dn_corr_A vg	W/m^2	Outgoing longwave radiation detected by the downward facing instrument, corrected	
Rs_net_Avg	W/m^2	Shortwave net radiation (Rshort_up - Rshort_down)	
Rl_net_Avg	W/m^2	Longwave net radiation (Rlong_up - Rlong_down)	
albedo_Avg	W/m^2	Albedo	
Rn_Avg	W/m^2	Net radiation (Rs_net + Rl_net)	
SQ_110_Avg	μmol photons $m^{-2}s^{-1}$	PAR (photosynthetically active radiation)	

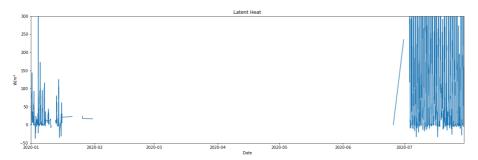
Variable Name	Units	Description	
shf_Avg(1)	W/m^2	Ground heat flux	
shf_Avg(2)	W/m^2	Ground heat flux	
slowsequence_ 3_Tot	samples	number of times scanned in 15 mins (once per min)	

Proposal

Background

Evapotranspiration (ET) is the process of water transferring from land to the atmosphere, accompanying the phase change of water from liquid to gas. This process plays a critical role in the ecohydrological system and profoundly affects the hydrological cycle. The processes of evapotranspiration and energy exchange are interdependent. Both latent heat (LE) and evapotranspiration (ET), from the perspective of energy and water flux, are key terms for anticipating weather conditions, simulating climate, and diagnosing climate change. However, the measurement of evapotranspiration is challenging because the process itself is invisible and complex.

Figure 2 shows the latent heat data gap in 2020 due to covid-19 and overhaul of equipment. Our project goal is to fill in these missing data. The ground truth data is collected from satelite sensors (https://etdata.org/). Despite the existence of numerous classical evapotranspiration simulation models, such as Bowen Ratio, Priestley-Taylor and Penman-Monteith models, the predictive accuracy of these models is inferior to that of deep learning models. Therefore, we plan to use RNN and LSTM deep learning models to predict latent heat and fill the gap.



Step 1: Regression analysis

We have 167 variables in the dataset. Although we can filter some ET related variables based on empirical models, these variables may not accurate and AI models tend to obtain adequate information. Therefore, we propose to conduct regression analysis to find out variables highly correlated to latent heat. These variables will be input variables in deep learning model.

Step 2: Deep Learning Time Series Forecast (Time Series Imputation)

Once we confirm the input variables, we plan to use RNN or LSTM forecast models to predict latent heat in 2020. All the input are divided into training datasets and the validation datasets. After the RNN

model is trained, the validation datasets are used to verify the model. At last, the missing data are generated by the model.

Source

This manuscript is a template (aka "rootstock") for <u>Manubot</u>, a tool for writing scholarly manuscripts. Use this template as a starting point for your manuscript.

The rest of this document is a full list of formatting elements/features supported by Manubot. Compare the input (.md files in the /content directory) to the output you see below.

Basic formatting

Bold text

Semi-bold text

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Right-aligned text

Italic text

Combined italics and bold

Strikethrough

- 1. Ordered list item
- 2. Ordered list item
 - a. Sub-item
 - b. Sub-item
 - i. Sub-sub-item
- 3. Ordered list item
 - a. Sub-item
- List item
- · List item
- · List item

subscript: H₂O is a liquid

superscript: 2¹⁰ is 1024.

unicode superscripts⁰¹²³⁴⁵⁶⁷⁸⁹

unicode subscriptS0123456789

A long paragraph of text. Lorem ipsum dolor sit amet, consectetur adipiscing elit, sed do eiusmod tempor incididunt ut labore et dolore magna aliqua. Ut enim ad minim veniam, quis nostrud exercitation ullamco laboris nisi ut aliquip ex ea commodo consequat. Duis aute irure dolor in reprehenderit in voluptate velit esse cillum dolore eu fugiat nulla pariatur. Excepteur sint occaecat cupidatat non proident, sunt in culpa qui officia deserunt mollit anim id est laborum.

Putting each sentence on its own line has numerous benefits with regard to <u>editing</u> and <u>version</u> <u>control</u>.

Line break without starting a new paragraph by putting two spaces at end of line.

Document organization

Document section headings:

Heading 1

Heading 2

Heading 3

Heading 4

Heading 5

Heading 6



Horizontal rule:

Heading 1's are recommended to be reserved for the title of the manuscript.

Heading 2's are recommended for broad sections such as Abstract, Methods, Conclusion, etc.

Heading 3's and Heading 4's are recommended for sub-sections.

Links

Bare URL link: https://manubot.org

<u>Long link with lots of words and stuff and junk and bleep and blah and stuff and other stuff and more stuff yeah</u>

Link with text

Link with hover text

Link by reference

Citations

Citation by DOI [1].

Citation by PubMed Central ID [2].

Citation by PubMed ID [3].

Citation by Wikidata ID [4].

Citation by ISBN [5].

Citation by URL [6].

Citation by alias [7].

Multiple citations can be put inside the same set of brackets [1,5,7]. Manubot plugins provide easier, more convenient visualization of and navigation between citations [2,3,7,8].

Citation tags (i.e. aliases) can be defined in their own paragraphs using Markdown's reference link syntax:

Referencing figures, tables, equations

Figure 1

Figure 2

```
Figure 3

Figure 4

Table 2

Equation 1

Equation 2
```

Quotes and code

Quoted text

Quoted block of text

Two roads diverged in a wood, and I—I took the one less traveled by, And that has made all the difference.

Code in the middle of normal text, aka inline code.

Code block with Python syntax highlighting:

```
from manubot.cite.doi import expand_short_doi

def test_expand_short_doi():
    doi = expand_short_doi("10/c3bp")
    # a string too long to fit within page:
    assert doi == "10.25313/2524-2695-2018-3-vliyanie-enhansera-copia-i-
        insulyatora-gypsy-na-sintez-ernk-modifikatsii-hromatina-i-
        svyazyvanie-insulyatornyh-belkov-vtransfetsirovannyh-geneticheskih-
        konstruktsiyah"
```

Code block with no syntax highlighting:

```
Exporting HTML manuscript
Exporting DOCX manuscript
Exporting PDF manuscript
```

Figures



Figure 1: A square image at actual size and with a bottom caption. Loaded from the latest version of image on GitHub.



Figure 2: An image too wide to fit within page at full size. Loaded from a specific (hashed) version of the image on GitHub.



Figure 3: A tall image with a specified height. Loaded from a specific (hashed) version of the image on GitHub.



Figure 4: A vector .svg image loaded from GitHub. The parameter sanitize=true is necessary to properly load SVGs hosted via GitHub URLs. White background specified to serve as a backdrop for transparent sections of the image. Note that if you want to export to Word (.docx), you need to download the image and reference it locally (e.g. content/images/vector.svg) instead of using a URL.

Tables

Table 2: A table with a top caption and specified relative column widths.

Bowling Scores	Jane	John	Alice	Bob
Game 1	150	187	210	105
Game 2	98	202	197	102
Game 3	123	180	238	134

Table 3: A table too wide to fit within page.

	Digits 1-33	Digits 34-66	Digits 67-99	Ref.
pi	3.14159265358979323 846264338327950	28841971693993751 0582097494459230	78164062862089986 2803482534211706	piday.org
е	2.71828182845904523 536028747135266	24977572470936999 5957496696762772	40766303535475945 7138217852516642	nasa.gov

Table 4: A table with merged cells using the attributes plugin.

	Colors		
Size	Text Color	Background Color	
big	blue	orange	
small	black	white	

Equations

A LaTeX equation:

$$\int_0^\infty e^{-x^2} dx = \frac{\sqrt{\pi}}{2} \tag{1}$$

An equation too long to fit within page:

$$x = a + b + c + d + e + f + g + h + i + j + k + l + m + n + o + p + q + r + s + t + u + v + w + x + y + z + 1 + 2 + 3 + 4 + 5 + 6 + 7 + 8 + 9$$
 (2)

Special

▲ WARNING The following features are only supported and intended for .html and .pdf exports. Journals are not likely to support them, and they may not display correctly when converted to other formats such as .docx.

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useful for general information - manubot.org

1 Blue Banner

useful for important information - manubot.org

♦ Light Red Banner useful for *warnings* - <u>manubot.org</u>

References

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Daniel S Himmelstein, Ariel Rodriguez Romero, Jacob G Levernier, Thomas Anthony Munro, Stephen Reid McLaughlin, Bastian Greshake Tzovaras, Casey S Greene *eLife* (2018-03-01) https://doi.org/ckcj

DOI: <u>10.7554/elife.32822</u> · PMID: <u>29424689</u> · PMCID: <u>PMC5832410</u>

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Douglas Heaven

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cOAlition S

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5. **Open access**

Peter Suber

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Journal of The Royal Society Interface (2018-04) https://doi.org/gddkhn

DOI: <u>10.1098/rsif.2017.0387</u> · PMID: <u>29618526</u> · PMCID: <u>PMC5938574</u>

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Daniel S Himmelstein, Vincent Rubinetti, David R Slochower, Dongbo Hu, Venkat S Malladi, Casey S Greene, Anthony Gitter

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DOI: 10.1371/journal.pcbi.1007128 · PMID: 31233491 · PMCID: PMC6611653