

# Flux Partitioning

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Background

# Motivation

NEE is net flux of two gross fluxes. The third post-processing step is partitioning the net flux (NEE) into its gross components  $GPP$  and  $R_{eco}$ .

$$NEE = R_{eco} - GPP$$

## Nighttime-Partitioning

estimate  $R_{eco} \sim T$  relationship of nighttime NEE (where  $GPP = 0$ )

## Daytime-Partitioning

fit a model of  $NEE$  to global radiation, VPD and temperature.

# Sort records to Daytime and Nighttime

The partitioning needs to distinguish carefully between night-time and day-time records.

classified as nighttime, if

1. Threshold of  $R_g < 10 \text{ Wm}^{-2}$
2. Daytime between compute times of sunrise and sunset

Nighttime flux partitioning

# Temperature sensitivity: $E_0$

Respiration is modelled by eq. of Lloyd & Tayler (1994)

$$R_{eco}(T; R_{Ref}, E_0) = R_{Ref} \exp \left[ E_0 \left( \frac{1}{T_{Ref} - T_0} - \frac{1}{T - T_0} \right) \right]$$

where  $T_0 = -46.02^\circ\text{C}$  and Reference temperature  $T_{Ref} = 15^\circ\text{C}$ .

Temperature sensitivity,  $E_0$ , is fitted to successive 15-day periods on trimmed data.

Annually aggregated,  $E_0$  is then the mean across valid estimates where 1) there were at least six records, 2) temperature ranged across at least  $5^\circ\text{C}$ , and 3) estimates were inside range of 30 to 450K.

# Respiration at reference temperature: $R_{Ref}$

Respiration at reference temperature,  $R_{Ref}$ , is re-estimated from nighttime data

- using the annual  $E_0$  temperature sensitivity estimate
- for 7-day windows shifted consecutively for 4 days.

Then its assigned to the central time-point of the 4-day period and linearly interpolated between periods.

Hence, the obtained respiration-temperature relationship varies across time.

# Gross fluxes: $R_{eco}$ and $GPP$

$$R_{eco} = f(T; R_{Ref}(t), E_0)$$

$$GPP = NEE - R_{eco}$$

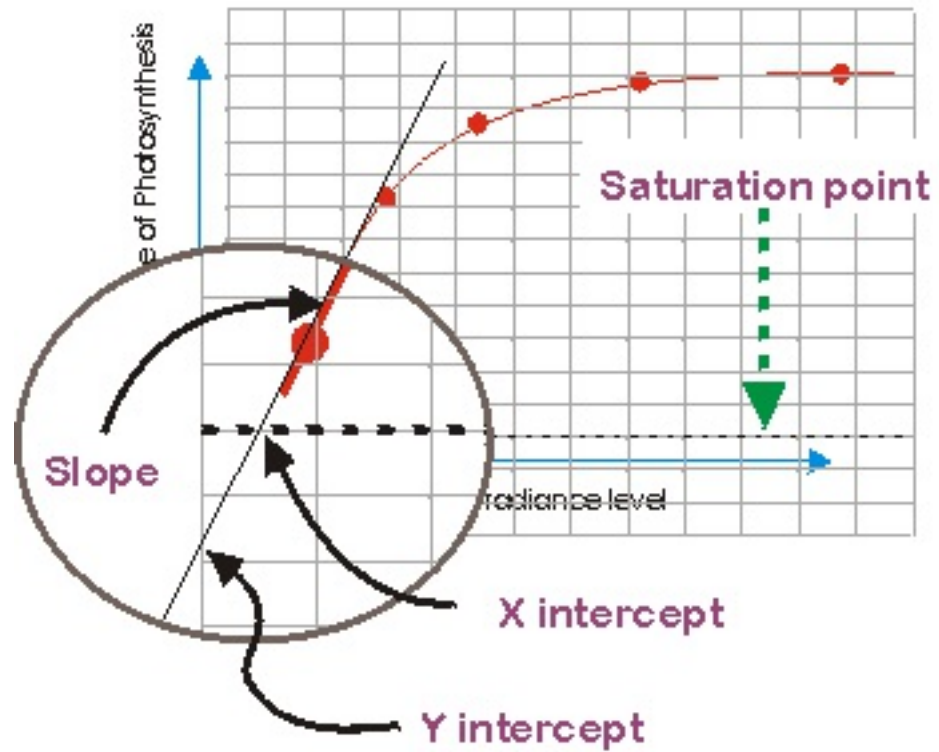
Note that variation in NEE can lead to negative  $GPP$ :

If predicted  $R_{eco} > NEE$  then predicted  $GPP < 0$ .



Daytime flux partitioning

# Light response curve (LRC)



# Fitting the Light response curve (LRC)

The method of Lasslop (2010) models  $NEE$ :

$$NEE = -GPP(R_g, VPD; \alpha, \beta_0, k) + R_{eco}(T; R_{Ref}, E_0)$$

$$GPP = \frac{\alpha \beta R_g}{\alpha R_g + \beta}$$

$$\beta = \begin{cases} \beta_0 \exp[-k(VPD - VPD_0)] & \text{if } VPD > 10 \text{ hPa} \\ \beta_0 & \text{otherwise} \end{cases}$$

$\alpha$  ( $\mu\text{mol CO}_2 \text{J}^{-1}$ ) is the canopy light utilization efficiency and represents the initial slope of the light-response curve,

$\beta$  ( $\mu\text{mol CO}_2 \text{m}^{-2} \text{s}^{-1}$ ) is the maximum CO<sub>2</sub> uptake rate of the canopy at infinite  $R_g$ , which is a decreasing function at higher VPD values.

# Temperature sensitivity

Temperature sensitivity,  $E_0$  is estimated from night-time data and provided to the day-time LRC fit to avoid parameter identifiability problems.

Different from the night-time partitioning a smoothed time varying estimate is used instead of the annual aggregate. And during  $E_0$  estimation, reference temperature  $R_{Ref}$  is set to the median temperature of the time window.

# LRC parameters and Reference temperature

**$R_{Ref}$**

Are are fitted using only daytime data and the previously determined temperature sensitivity ( $E_0$ ) for each shifting window across records.

# Gross fluxes: $R_{eco}$ and $GPP$

Are predicted by the LRC and Lloyd & Taylor respiration for each central record of the shifting window.

Results are linearly interpolated by the difference to the window centers.

## Lasslop 2010

Daytime estimates of reference temperatures are also used for predicting nighttime  $R_{eco}$

## Keenan 2019

Nighttime estimates of reference temperatures (obtained with the  $E_0$  fits) are used for predicting nighttime  $R_{eco}$

# Caution Partitioning is not always applicable

Partitioning only works if there is a good  $R_{eco} \sim T$  relationship.

It is not applicable if either there is

- Suppressed respiration at freezing temperatures
- Limited variation of temperature
- Strong controls of other factors such as moisture

# Flux Partitioning in REddyProc



# Preparations

Specify geographical coordinates and time zone.

Fill missing values in the used meteorological data.

```
EProc$setLocationInfo(LatDeg = 51.0, LongDeg = 13.6, TimeZoneHour = 1)
EProc$MDSGapFill('Tair', FillAll = FALSE, minNWarnRunLength = NA)
EProc$MDSGapFill('VPD', FillAll = FALSE, minNWarnRunLength = NA)
EProc$MDSGapFill('Rg', FillAll = FALSE, minNWarnRunLength = NA)
```

# Nighttime partitioning

Repeated for each of the  $u_*$  threshold scenario (percentiles of  $u_* Th$  distribution)

```
EProc$MRFluxPartitionUStarScens()  
grep("GPP|Reco", names(EProc$ExportResults()), value = TRUE)
```

```
## [1] "Reco_uStar"      "GPP_uStar_f"     "GPP_uStar_fqc"  "Reco_U05"  
## [5] "GPP_U05_f"       "GPP_U05_fqc"     "Reco_U50"       "GPP_U50_f"  
## [9] "GPP_U50_fqc"     "Reco_U95"        "GPP_U95_f"      "GPP_U95_fqc"
```

It produces output columns **Reco\_<uStar>** and **GPP\_<uStar>\_f** modified by the respective  $u_*$  threshold suffix in the REddyProc class.

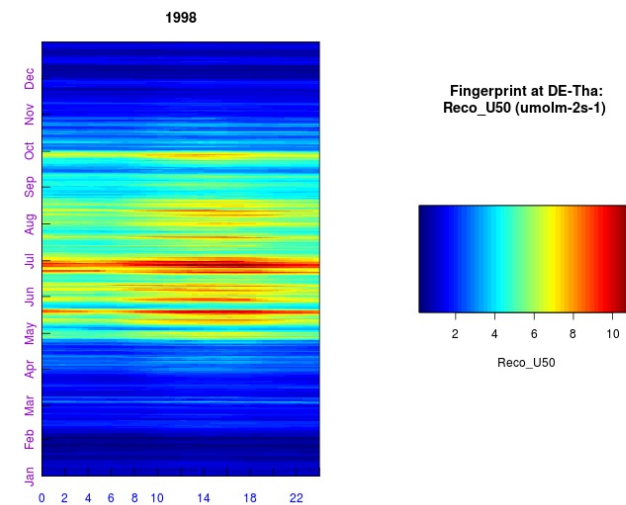
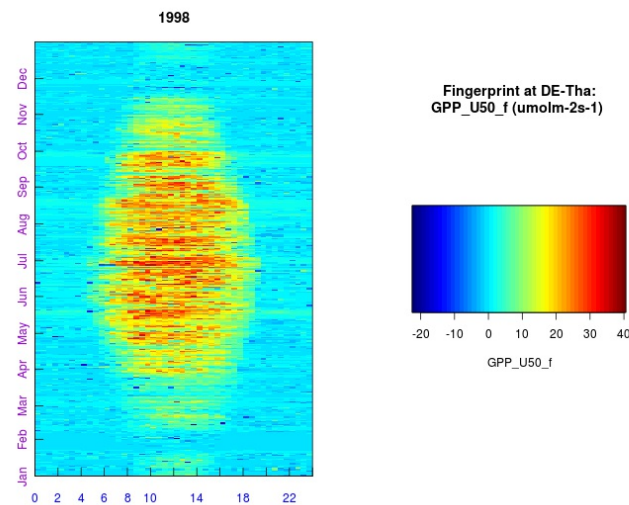
**GPP\_<uStar>\_fqc > 1** denotes bad quality for windows where valid parameter estimates are further away.

# Fingerprint plots of Nighttime $R_{eco}$ and $GPP$

```
EProc$PlotFingerprint('GPP_U50_f', Dir = "plots", Format = "png")  
EProc$PlotFingerprint('Reco_U50', Dir = "plots", Format = "png")
```

*GPP*

$R_{eco}$



# Daytime partitioning

Repeated for each of the  $u_*$  threshold scenario (percentiles of  $u_* Th$  distribution)

```
invisible(EProc$GLFluxPartitionUStarScens())  
grep("GPP.*_DT|Reco.*_DT", names(EProc$ExportResults()), value = TRUE)
```

```
## [1] "Reco_DT_uStar"      "GPP_DT_uStar"      "Reco_DT_uStar_SD"  
## [4] "GPP_DT_uStar_SD"    "Reco_DT_U05"       "GPP_DT_U05"  
## [7] "Reco_DT_U05_SD"     "GPP_DT_U05_SD"     "Reco_DT_U50"  
## [10] "GPP_DT_U50"         "Reco_DT_U50_SD"    "GPP_DT_U50_SD"  
## [13] "Reco_DT_U95"        "GPP_DT_U95"        "Reco_DT_U95_SD"  
## [16] "GPP_DT_U95_SD"
```

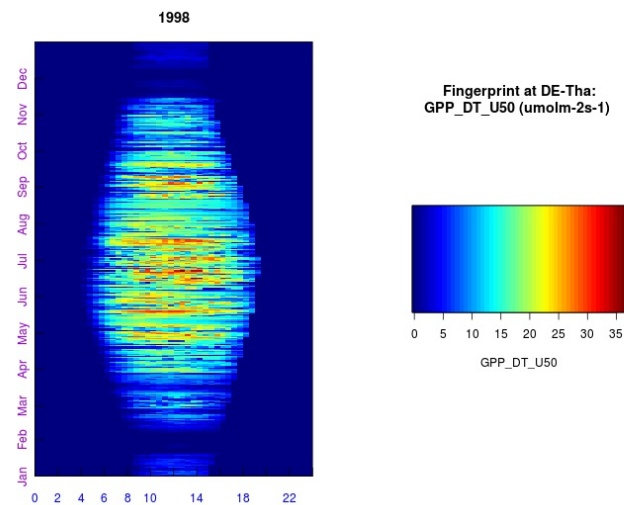
It produces output columns **Reco\_DT\_<uStar>** and **GPP\_DT\_<uStar>** modified by the respective  $u_*$  threshold suffix in the REddyProc class.

**\_SD** denotes the standard deviation of the prediction error.

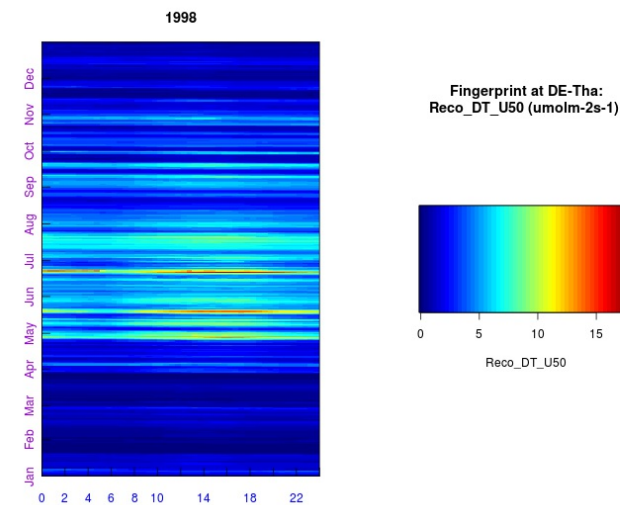
# Fingerprint plots of Daytime $R_{eco}$ and $GPP$

```
EProc$PlotFingerprint('GPP_DT_U50', Dir = "plots", Format = "png")  
EProc$PlotFingerprint('Reco_DT_U50', Dir = "plots", Format = "png")
```

$GPP$



$R_{eco}$



# Save the results

So far the results are stored internally in the REddyProc class. Get them as a dataframe:

```
results = EProc$ExportResults()  
# bind to original original data  
appResults = cbind(EddyData, results)  
# save to tab-separated file  
fWriteDataframeToFile(appResults, "DETha98_proc.txt")
```

```
## Number of NA converted to '-9999': 615607
```

```
## Wrote tab separated textfile: DETha98_proc.txt
```