



Estimating Evapotranspiration using Combined Physics-Based and Data-Driven Machine Learning

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*Evapotranspiration* as a Key Flux: Looking into *Stomatal* and *Aerodynamic Resistance* 





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## Hybrid Model

## ТЛП



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Multi-Task Learning

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A-priori Information

**Big Leaf** 

Model



## The FLUXNET 2015 dataset

Site ID	IGBP	Elevation (m)	Mean Annual Temperature (°C)	Mean Annual Precipitation (mm)	Data Availability
DE-Tha	ENF	385	8.2	843	19 years (1996 - 2014)
FR-Pue	EBF	270	13.5	883	15 years (2000 - 2014)
FR-LBr	ENF	61	13.6	900	12 years (1996 - 2008)
CH-Cha	GRA	393	9.5	1136	10 YEARS (2005 - 2014)
DE-Gri	GRA	385	7.8	901	11 YEARS (2004 - 2014)
US-Var	GRA	129	15.8	559	15 YEARS (2000 - 2014)

•ENF (Evergreen Needleleaf Forests: Lands dominated by woody vegetation with a percent cover >60% and height exceeding 2 meters. Almost all trees remain green all year. Canopy is never without green foliage).

•EBF (Evergreen Broadleaf Forests: Lands dominated by woody vegetation with a percent cover >60% and height exceeding 2 meters. Almost all trees and shrubs remain green yearround. Canopy is never without green foliage).

•GRA (Grasslands: Lands with herbaceous types of cover. Tree and shrub cover is less than 10%. Permanent wetlands lands with a permanent mixture of water and herbaceous or woody vegetation. The vegetation can be present in either salt, brackish, or fresh water.)



## Evaluating *predictions* against *observations*





Evaluating *latent heat flux* predictions against relevant *climate variables* 





Evaluating *predictions* against relevant *climate variables* 



Predicted latent variable evaluated against constrained model	R <sup>2</sup> metric value averaged over sites	
r latant variable	a priori constraint	$R^2 = 0.901$
	Multi-Task Learning	$R^2 = 0.826$
r latent veniable	a priori constraint	$R^2 = 0.055$
	Multi-Task Learning	$R^2 = 0.261$

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Conclusion



- Hybrid modelling allows
  - the retrieval of *latent variables*
  - that are *physically interpretable* in comparison to both *purely data-driven* and *physicsbased approaches,* and
- *Equifinality* can be circumvented by inducing more theory or data
- Data-driven modelling *needs* to be boosted by domain expertise!