

performance of arable annual mixed crops

Rémi Mahmoud, Noémie Gaudio, Pierre Casadebaig, Xavier Gendre, Laurent Bedoussac, Guénaëlle Corre-Hellou, Florian Fort, Etienne-Pascal Journet, Isabelle Litrico, Christophe Naudin, Cyrille Violle



Agronomic context: towards a more sustainable agriculture

Intensive agriculture

Environment optimized through external inputs

Agroecological transition
System diversification
(field, rotation, landscape)

Sustainable agriculture

Less or no inputs → limiting conditions
Use of plant diversification

Agronomic context: towards a more sustainable agriculture

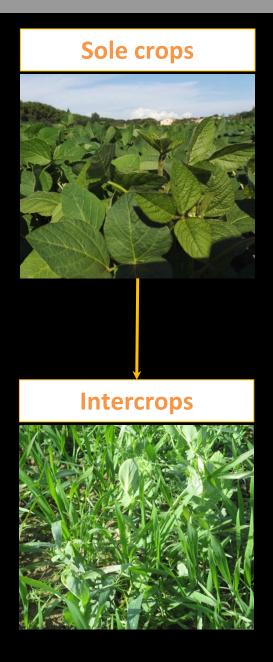
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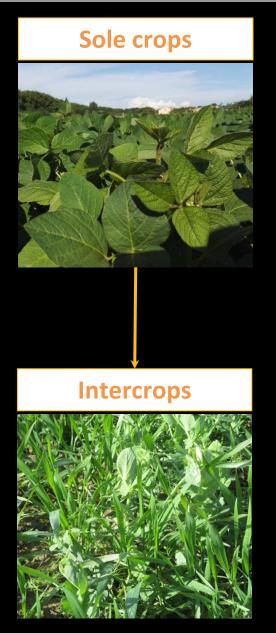
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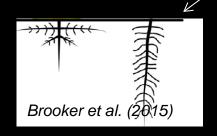
Less or no inputs → limiting conditions
Use of plant diversification

Maximize complementarities and positive interactions



Conceptual framework: Recipes of ecology to make efficient mixtures

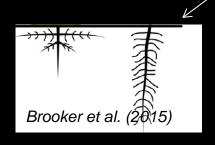
Divergence between species Complementarity or competition



Metrics = distance for target functional traits between the two species



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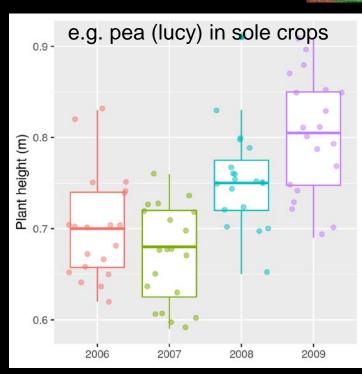
Metrics = distance for target functional traits between the two species



Phenotypic plasticity

- 1 genotype \leftrightarrow *n* phenotypes
- = Adaptation to the environment

Metrics = variance of a target trait between environments



Reviewed by Nicotra et al (2010) Trends Plant Sci 15:684-692 Conceptual framework: Recipes of ecology to make efficient mixtures

What about the transposability of these concepts to agroecosystems?

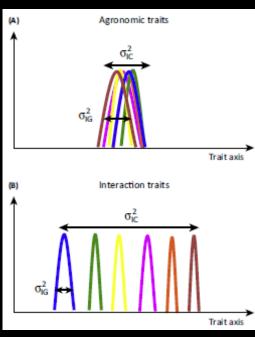
- Low diversity compared to natural communities
- Artificial selection for sole crops rather than natural selection

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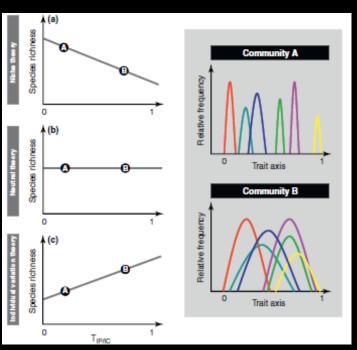
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- Low diversity compared to natural communities
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Can we gather relevant data to add experimental points on theoretical curves?



Litrico I, Violle C (2015) Trends Plant Sci. 20:604–613



Violle C et al (2012)
Trends in Ecology & Evolution 27:244–252

Objective

- Agronomic interest of intercrops experimentally demonstrated
- High variability of results according to environmental conditions

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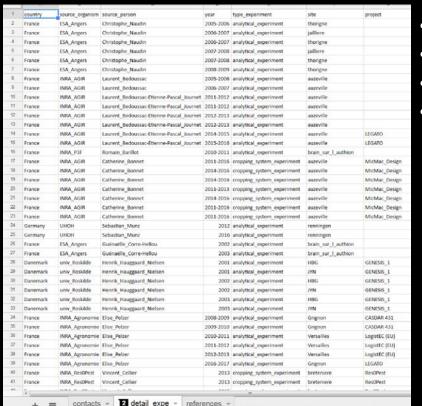
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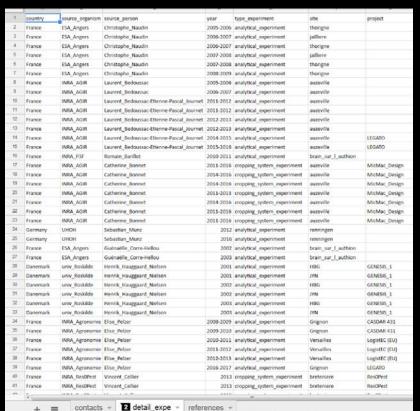


Objective: develop statistical models to predict intercrops performance based on environmental conditions (soil, climate, practices), species intrinsic differences (trait values) and species plasticity (trait variance).

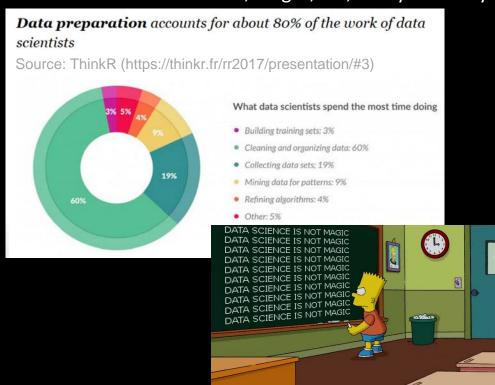


- France, Germany, UK, Denmark, Netherlands, Italy
- 42 field experimentations, in 15 sites and 16 years
- 153 intercrops combinations and ~ 10.000 plots
- Measured traits: biomass, height, LAI, N...dynamically

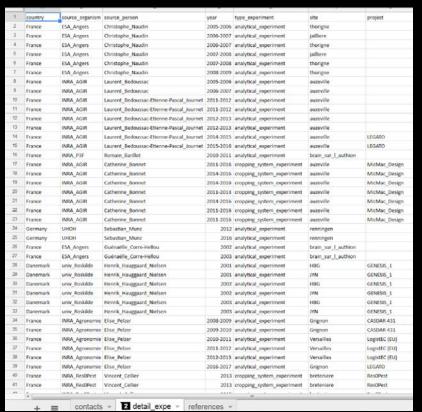
Data-driven models (empirical)



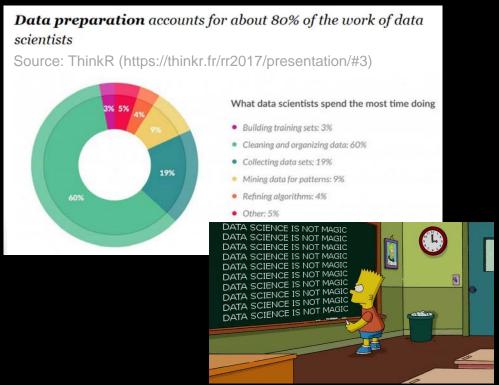
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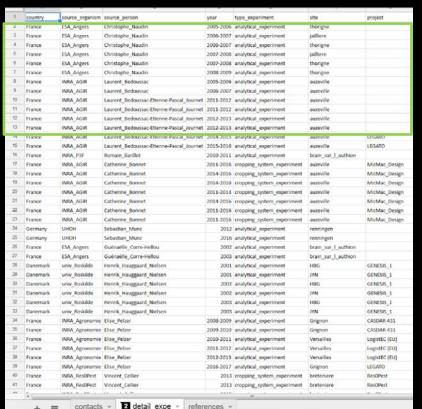
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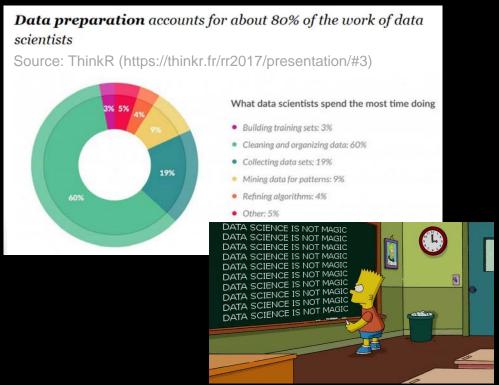
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"intelligent": driven by ecological concepts

= mobilize agronomic and ecological knowledge to define integrative variables that are considered important for the performance of the modeled system



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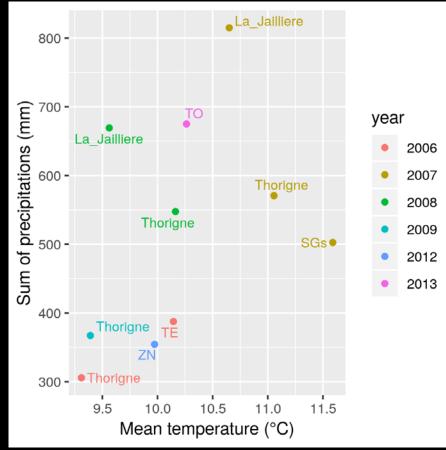
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Preliminary results on a data subset

		-			•		
1	country	source_organism	source_person	year	type_experiment	site	project
Z	France	ESA_Angers	Christophe_Naudin	2005-2006	analytical_experiment	thorigne	
3	France	ESA_Angers	Christophe_Naudin	2006-2007	analytical_experiment	jailliere	
4	France	ESA_Angers	Christophe_Naudin	2006-2007	analytical_experiment	thorigne	
5	France	ESA_Angers	Christophe_Naudin	2007-2008	analytical_experiment	jailliere	
6	France	ESA_Angers	Christophe_Naudin	2007-2008	analytical_experiment	thorigne	
7	France	ESA_Angers	Christophe_Naudin	2008-2009	analytical_experiment	thorigne	
8	France	INRA_AGIR	Laurent_Bedoussac	2005-2006	analytical_experiment	auzeville	
9	France	INRA_AGIR	Laurent_Bedoussac	2006-2007	analytical_experiment	auzeville	
10	France.	INRA_AGIR	Laurent_Bedoussac-Etienne-Pascal_Journet	2011-2012	analytical_experiment	auzeville	
11	France	INRA_AGIR	Laurent_Bedoussac-Etienne-Pascal_Journet	2011-2012	analytical_experiment	auzeville	
12	France	INRA_AGIR	Laurent_Bedoussac-Etienne-Pascal_Journet	2012-2013	analytical_experiment	auzeville	
13	France	INRA_AGIR	Laurent_Bedoussac-Etienne-Pascal_Journet	2012-2013	analytical_experiment	auzeville	
14	France	INRA_AGIR	Laurent_Bedoussac-Etienne-Pascal_Journet	2014-2015	analytical_experiment	auzeville	LEGATO
15	France	INRA_AGIR	Laurent_Bedoussac-Etienne-Pascal_Journet	2015-2016	analytical_experiment	auzeville	LEGATO
16	France	INRA_P3F	Romain_Barillot	2010-2011	analytical_experiment	brain_sur_l_authion	
17	France	INRA_AGIR	Catherine_Bonnet	2011-2016	cropping_system_experiment	auzeville	MicMac_Design
18	France	INRA_AGIR	Catherine_Bonnet	2014-2016	cropping_system_experiment	auzeville	MicMac_Design
19	France	INRA AGIR	Catherine_Bonnet	2014-2016	cropping system experiment	auzeville	MicMac Design
20	France	INRA_AGIR	Catherine_Bonnet	2011-2013	cropping_system_experiment	auzeville	MicMac_Design
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22	France	INRA AGIR	Catherine Bonnet	2011-2016	cropping system experiment	auzeville	MicMac Design
23	France	INRA AGIR	Catherine Bonnet	2011-2016	cropping system experiment	auzeville	MicMac Design
24	Germany	UHOH	Sebastian Munz	2012	analytical experiment	renningen	
25	Germany	UHOH	Sebastian Munz	2016	analytical_experiment	renningen	
26	France	ESA Angers	Guénaëlle Corre-Hellou	2002	analytical_experiment	brain_sur_I authion	
27	France	ESA Angers	Guénaëlle Corre-Hellou	2003	analytical experiment	brain sur I authion	
28	Danemark	univ Roskilde	Henrik Hauzgaard Nielsen	2001	analytical experiment	HBG	GENESIS 1
29	Danemark	univ_Roskilde	Henrik Hauggaard Nielsen	2001	analytical experiment	JYN	GENESIS 1
30	Danemark	univ Roskilde	Henrik Hauggaard Nielsen	2002	analytical experiment	HBG	GENESIS 1
31	Danemark	univ Roskilde	Henrik Hauggaard Nielsen	2002	analytical experiment	JYN	GENESIS 1
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33	Danemark	univ Roskilde	Henrik Hauggaard Nielsen	2003		IYN	GENESIS_1
34	France	INRA Agronomie		2008-2009		Grignon	CASDAR 431
35	France	INRA Agronomie		2009-2010		Grignon	CASDAR 431
36	France	INRA Agronomie		2010-2011		Versailles	LogistEC (EU)
37	France	INRA Agronomie		2011-2012		Versuilles	LogistEC (EU)
38	France	INRA Agronomie		2012-2013		Versailles	LogistEC (EU)
39	France	INRA Agronomie		2016-2017		Grignon	LEGATO
40	France	INRA ResOPest	Vincent Cellier		cropping system experiment	breteniere	ResOPest
41	France	INRA ResOPest	Vincent Cellier		cropping system experiment	breteniere	ResOPest
47		INKA_Resurest	Vincent Celler	2013		breteriere	Resurest
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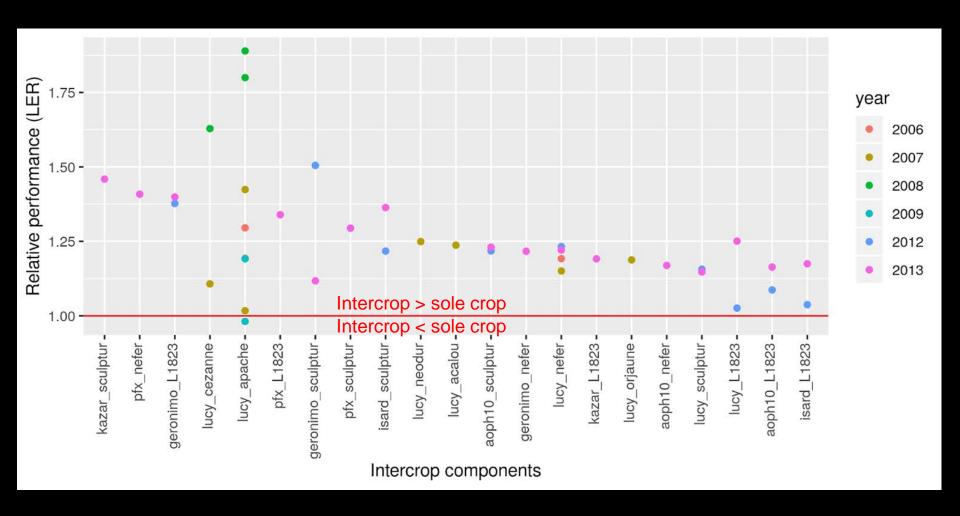
Data subset:

- 10 experiments in France (two sites: INRA AGIR, ESA Angers)
- 6 years
- 40 intercrops combinations of wheat and pea



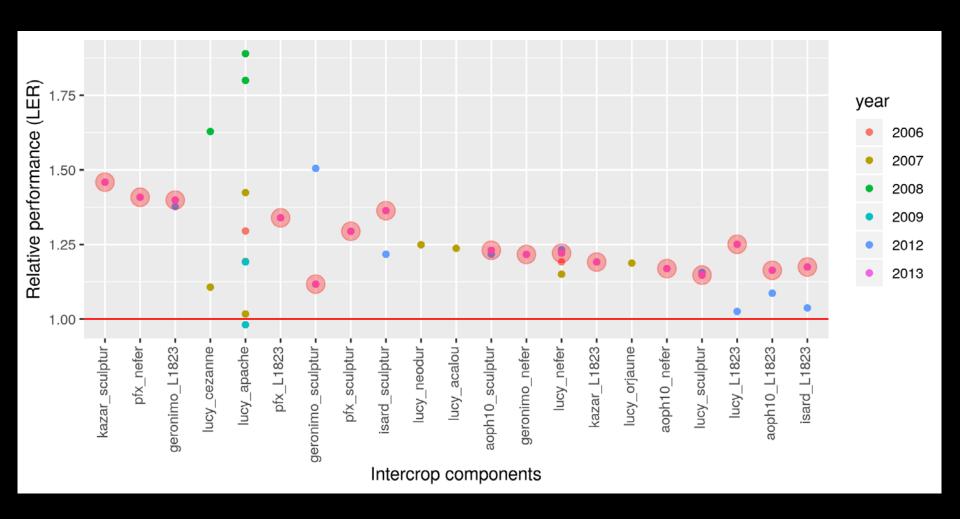
Rémi Mahmoud (internship Master 2, 2018)

Successful but variable performance of intercrops in low input contexts



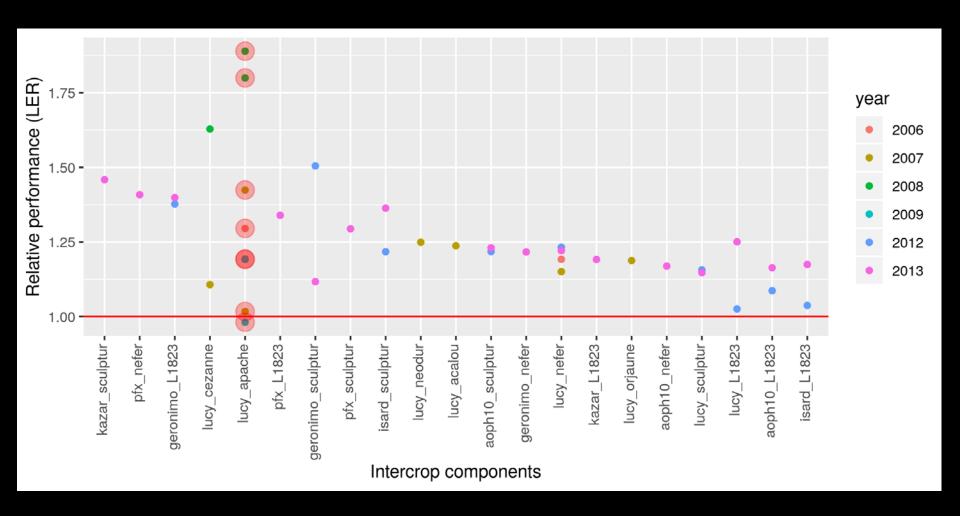
$$LER_{trait} = \frac{trait_{ICi}}{trait_{SCi}} + \frac{trait_{ICj}}{trait_{SCj}}$$

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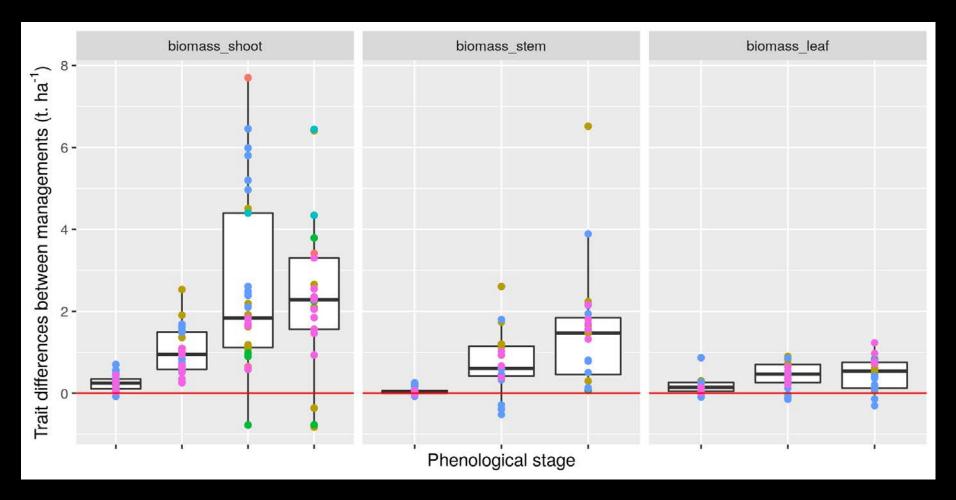


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Preliminary results on a data subset

Effect of mixtures on species characteristics (plasticity)

Species adapt strongly when growing in intercrop (illustration for wheat)

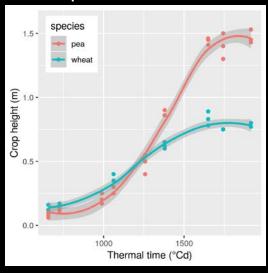


 $\Delta trait_{(intercrop, sole\ crop)} = trait_{intercrop} - trait_{sole\ crop}$

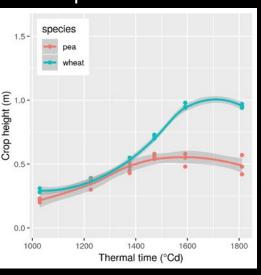
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Differences between species within intercrops (inter-specific divergence)



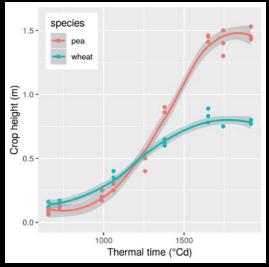


pea < wheat

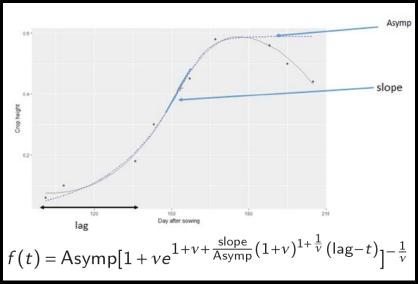


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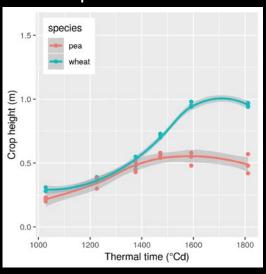




curve parameters



pea < wheat

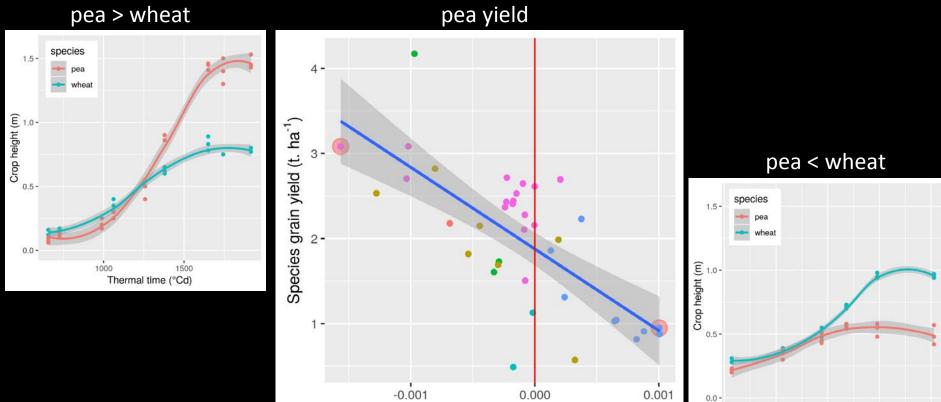


 $\Delta trait_{(wheat,pea)} = trait_{wheat} - trait_{pea}$

Preliminary results on a data subset

Performance of a species according to the difference between species

In intercrop, pea performance is linked to growth strategies (predictors)



Thermal time (°Cd)

 $Yield_{pea} = f(growth \, rate_{wheat} - growth \, rate_{pea})$

Difference between species (elongation, m. Cd⁻¹)

Perspectives

Still a lot of work

Identify other variables linked to the cover functioning + other mixtures

→ Develop the models, from multiple regression to machine learning (decision tree, random forests, deep learning)

Difficulties identified

The traits measured in agronomic experiments do not systematically reflect niche complementarity (height, biomass)

Take into account phenotypic plasticity to build generic assembly rules for different cropping conditions

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For the photographs, thanks to G Corre-Hellou, J Evers (univ. Wageningen), C Bonnet (INRA), L Bedoussac