

The European Commission's science and knowledge service

Joint Research Centre

Estimation of crop residue production in Europe with empirical models

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European
Commission

Context

Objective, assessment approach, state of art, knowledge gaps

Objective: quantify and spatialize the current biomass production from agriculture in Europe without considering final uses

Biomass (BY)

Y : available statistics

R : PART TO
ESTIMATE

Approach for assessment:

- Infer crop **residues (R)** from crop **yield (Y)** statistics: used in large-extent assessments of biomass potentials.

- Through coefficients (biomass partitioning) like the **harvest index (HI)** or the residue-to-product ratio (**RPR**)

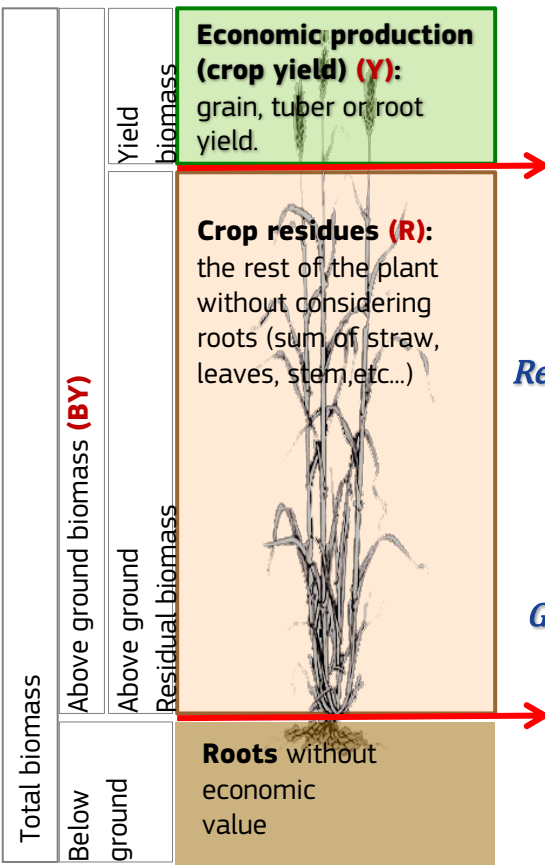
$$HI = \frac{Y}{Y + R} = \frac{Y}{BY} \quad RPR = \frac{R}{Y}$$

Review of similar studies :

- Constant crop-specific **HI** coefficient (or **RPR**).
- Empirical regression models (e.g Log,Exp), assuming + correlation **RPR(HI)** and **Y** (e.g. Scarlat et al. 2010, Bentsen et al. 2014).

Going further...faced questions:

- Do **positive relationships** between **HI** and **Y** exist? Is the nature of that relationship **similar for all crops**? Agronomic **reasons** behind this?
- How much **uncertainty** can we expect when estimating **crop residues** using such empirical models?
- How can model estimations and uncertainties **influence** the assessment of residues production **at EU28 level**?



Methodology – Part 1: Creating empirical models

Collection of experimental data from scientific literature

Extensive literature review of papers in English reporting **experimental data on economic yield and biomass partitioning** (*HI*, harvest index)

From a first selection of about **120 papers**, extracting **1580 observations** based on their definition of *HI*.

Yields were **transformed to dry-matter content**

Number of observations used per geographical area

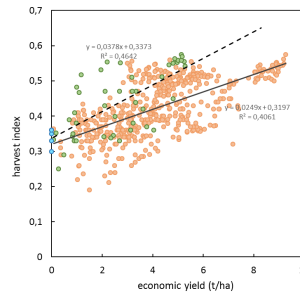
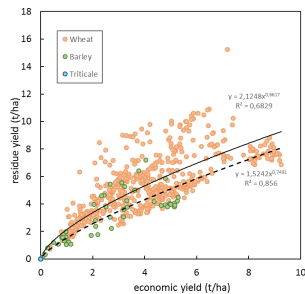
Geographical region\Crop	wheat	barley	maize	rice	sorghum	rapeseed	soybean	sunflower	potato	sugar beet	Total Regions
EU-28	200	13	23		6	82	12	4	40	19	399
Europe (others)	10										10
North America	86	12	115		15	8	54		6		296
South America			46						18		64
Middle East	36	12	16			100		34		26	224
Southern Africa	9										9
Southern Asia	56		45	20	69			32			222
South-eastern Asia				59			29				88
Eastern Asia	40			52		34	18		8		152
Australia	66	4			38						116
Total Crops	503	41	245	131	128	224	113	78	72	45	1580

Methodology – Part 1: Creating empirical models

1) FINDINGS ABOUT: Nature of the relationship between Y and HI

Groups of crops with 2 different behaviours in the relationship Y - HI

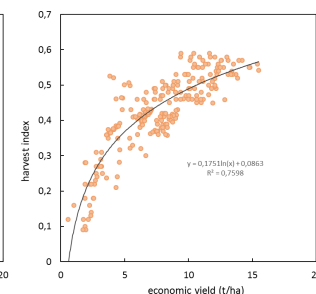
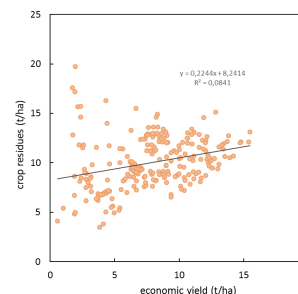
Wheat, barley, rice, sunflower, rapeseed



- Y and R are strongly correlated
- Consequently, HI is stable, and weakly correlated with Y

In these crops, kernel number is determined during vegetative growth (tillering, heading phases), which makes final yield correlated with vegetative biomass

Maize, sorghum, soybean, potato and sugar beet



- Y and R are poorly correlated
- Consequently, HI is variable, and strongly correlated with Y

In maize and sorghum, both kernel number and grain weight are after the plant vegetative growth

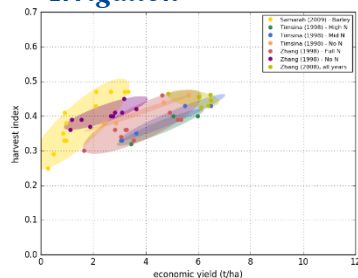
Potato and sugarbeet vegetative biomass is detached from yield

2) FINDINGS ABOUT: How Y and biomass partitioning (HI) are affected by the influence of environmental/genetics/management

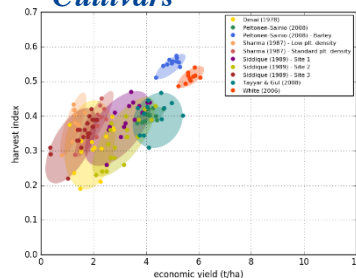
Irrigation (water availability) introduces significant changes in both HI and Y

Genetics introduce variability in HI but not too much in Y . N fertilizing changes Y , but not the HI

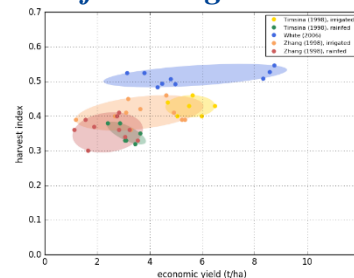
Irrigation



Cultivars



N fertilizing



Therefore, empirical models between Y and HI are mainly describing changes in water availability

Methodology – Part 1: Creating empirical models

3) Producing new empirical models to predict R from Y with 95% confidence intervals

FINDINGS

Statistical analysis to construct robust regression models

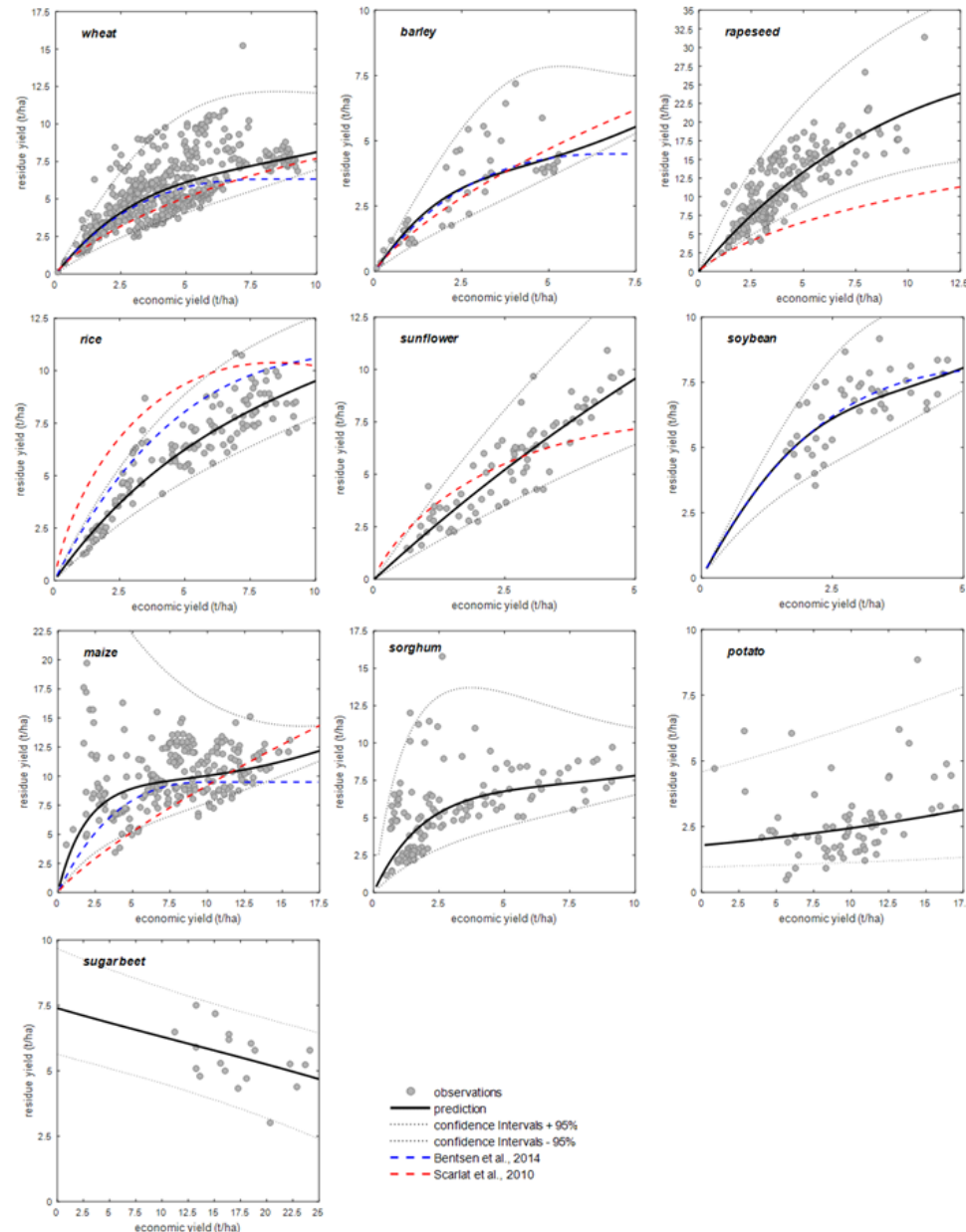
- **Finding:** Need of using *HI* as predicted variable to remove heteroscedasticity (R-Y) when computing models.
- Apply transformations for normality of residuals

Model **uncertainties** are **quite large** (e.g maize). Mostly due to differences in the crop varieties and other management factors (N fertilizing)

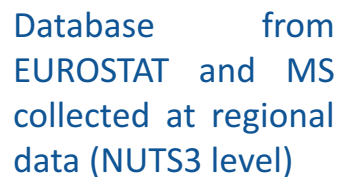
Overall agreement with **other studies** in the **main crops** (wheat, barley, maize).

- Sometimes the differences against Bentsen et al. (2014) or Scarlat (2010) are due to the models imposed (e.g. logarithmic, exponential)

Crops modelled cover 98% of EU28 crop residue production.



Implementation of the models within EU28 and estimation for the minority crops



Post-processing
algorithm to
generate a complete
dataset

Estimation in dry matter at NUTS3 level

Economic production

Models: Cereals, oilseeds, sugar and starchy crops. (98% EU28 RP)
Fixed coefficients: permanent crops, (others: pulses, industrial)

Vegetables, fodder crops, energy crops

Disaggregation 25 km²

Land cover data (CLC)

$$B_g = \sum_i^n B_{c,i,N=3} * \frac{L_{g \cap i}}{L_{i,N=3}}$$

Results – Big numbers (Dry matter Mt)

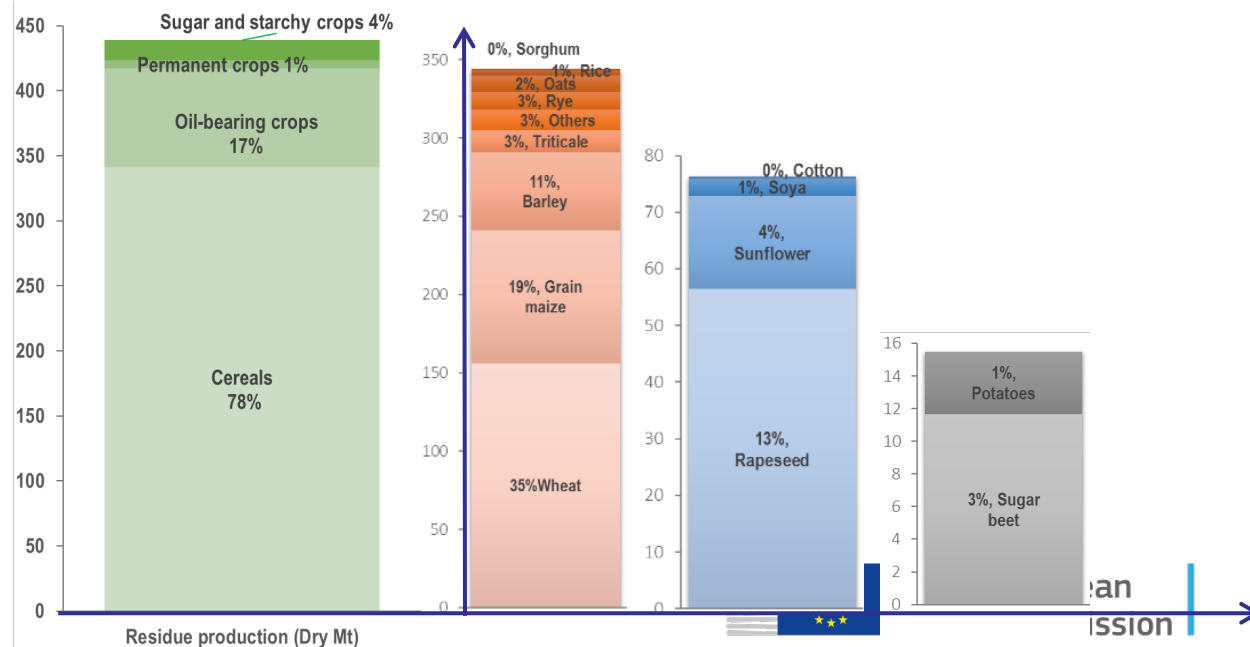
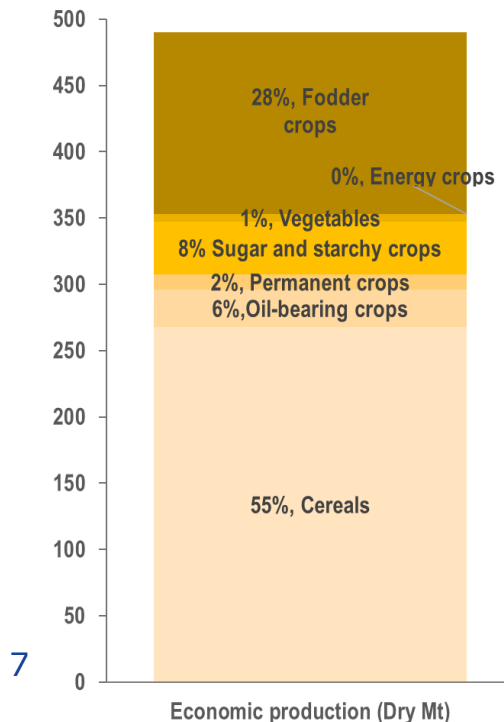
Crop group	Total	Economic	Residue	Upper	Lower
	Biomass	production	production	CI	CI
Cereals	609	268	341	643	227
Energy crops	0.19	0.19			
Fodder crops	137	137			
Oil-bearing crops	104	28	76	131	50
Permanent crops (+ others)	18	12	6		
Sugar and starchy crops	55	40	15	22	13
Vegetables	6	6			
Total	929	490	439	796	290

Biomass production

**Economic production :
(Y)**

**Residue production :
(R)**

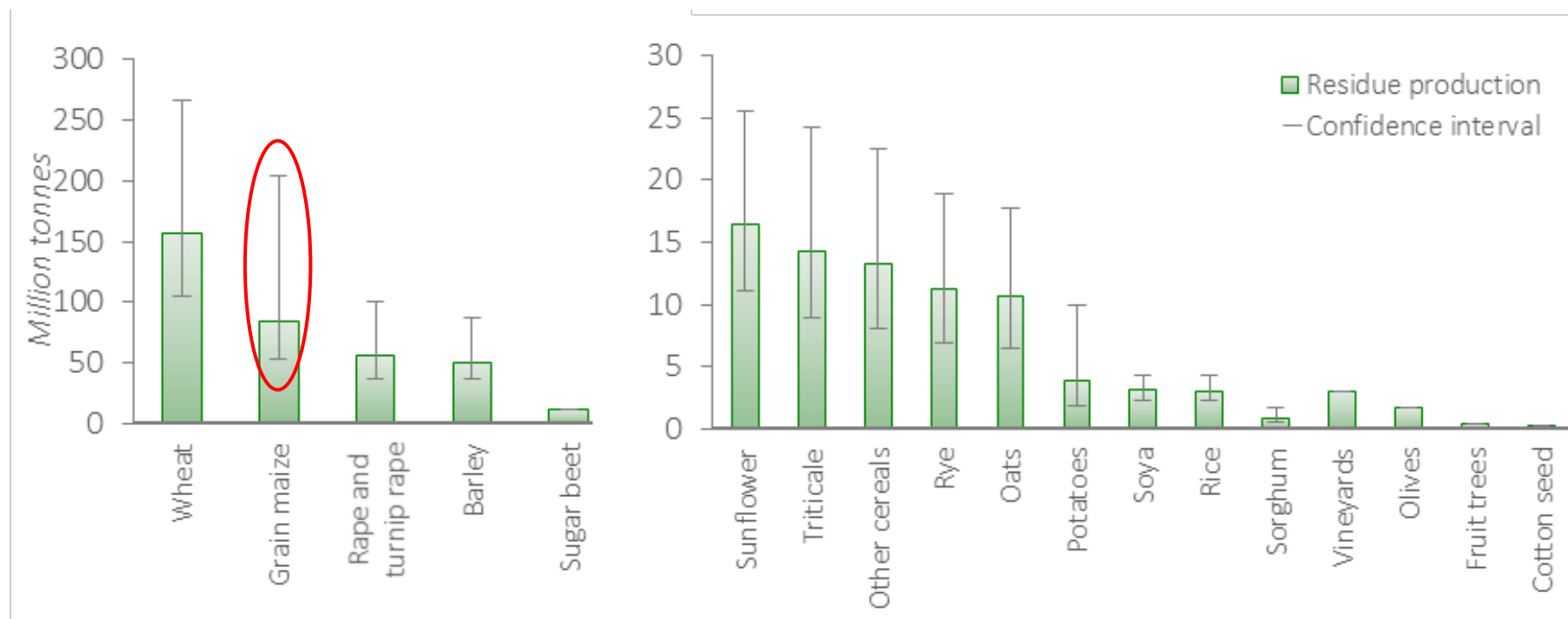
Average (2011-2015)



Results – Uncertainties of estimations by crops

The **uncertainties of the estimations in EU28 are high**, especially for maize, the second in importance

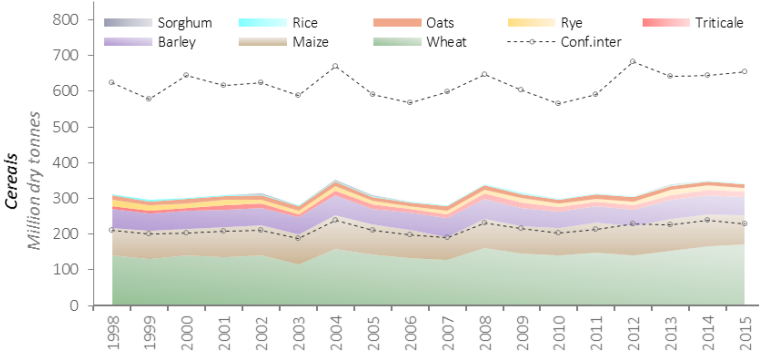
Mostly due to differences in the **crop varieties** and other management factors (**N fertilizing**) in the **data collected** since they are coming from experimental conditions.



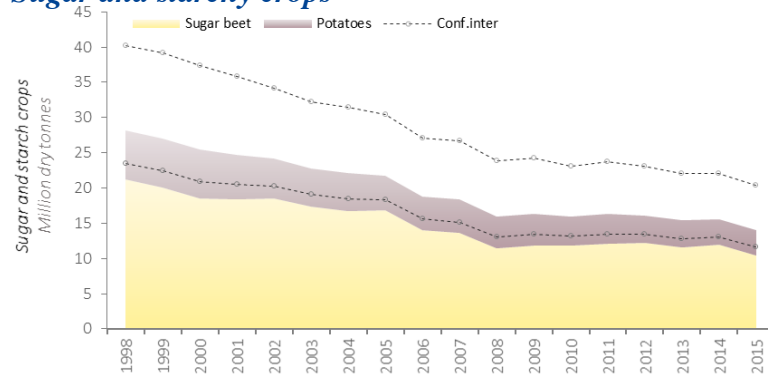
Results – Evolution and distribution in Europe

Evolution of crop production (1998-2015)

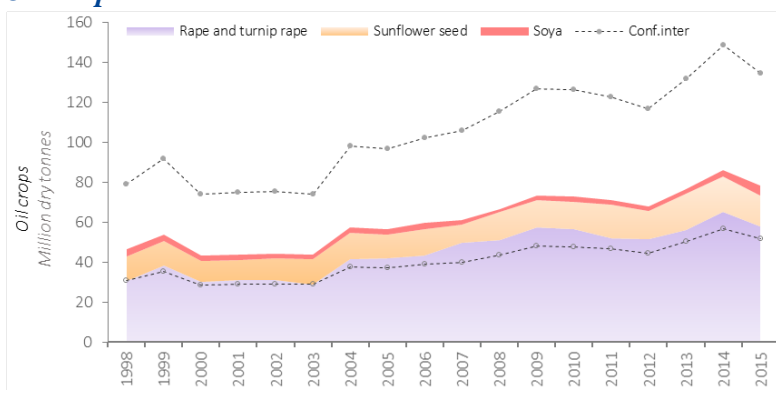
Cereals



Sugar and starchy crops

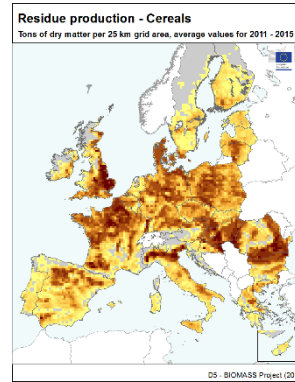
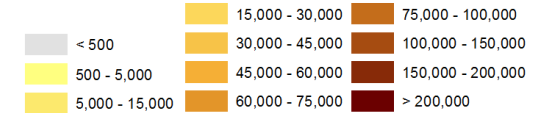


Oil crops



Current estimations (avg 2011-2015)

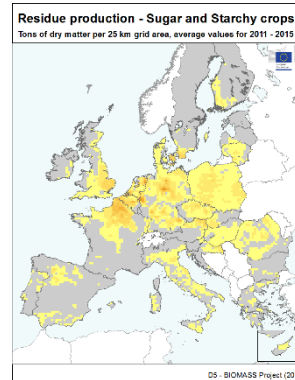
Tons of dry matter per 25 km grid per year



Stable production

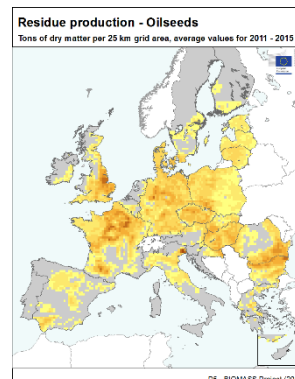
Production of residues **evenly distributed** across EU28

Inter-annual **variability** is driven by **weather**



Decrease of residue production, driven by a progressive **reduction of areas** (e.g. impact of EU-policies in the case of sugar beet (EU sugar production quotas))

Production mainly concentrated:
FR, DE, UK



Progressive augmentation, driven by an increase of **rapeseed area**. Consequence of the **increasing use as biofuel** (e.g. after biofuel EU-policy).

Conclusions

A relationship between Y and HI exists in the crops studied **BUT**:

- It varies significantly depending on the crop, and is mainly describing effects of water availability.
- A regression between Y and HI seems a priori of little use: **when HI is correlated to yield Y , R tends to be constant.**
- **HI is only useful** as predicted variable **to solve problems of heteroscedasticity.**

Empirical models produce **high uncertainties**, and biophysical models (EO data) are needed to reduce these **large uncertainties observed**.

Residues production in EU28 is estimated at 439 dry Mt/year, with four crops (wheat, maize, barley, rapeseed) accounting for 80% of this.

Some crops (e.g. sugar beet, rapeseed) **present an appreciable production trend** due to gradual changes in sown area, partially reflecting the effect of EU policies.

Thank you



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