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Connecting Variety Trialling Systems Across Two Countries

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Introduction

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- This testing for VCU is done by examination offices (EO) at the national level in many European countries.
- Ususally, VCU trials last 2 to 3 years and involve 5 to 15 locations per year.

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- Moreover, it may be marketed in other European countries as well, even if it has not been tested for VCU in that country.
- Up to that point, no performance assessment has been conducted in those other countries.

Objective

- We investigate how VCU trialling systems of two countries can be connected and integrated.

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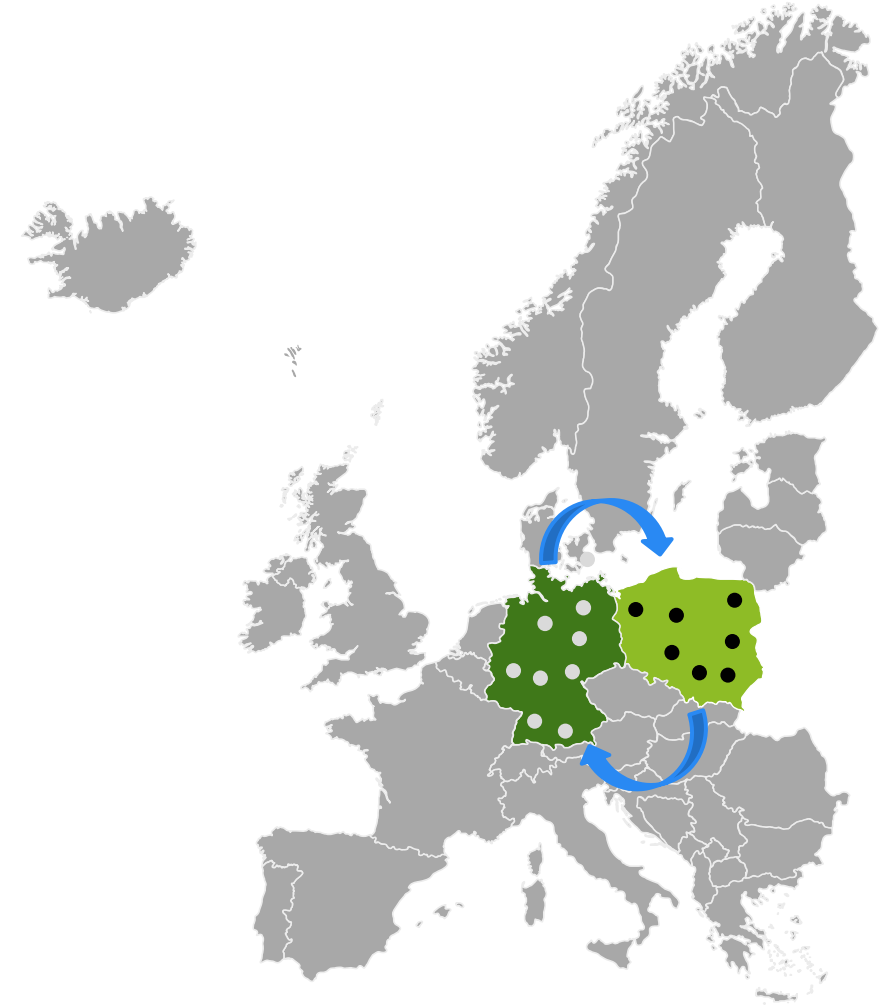
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- So that performance can be assessed for both countries without altering the capacity of the trialling systems.
- We highlight the benefit of a joint analysis of data from two countries, and show how the efficiency of each country's VCU testing system can be improved—without altering the overall capacity of the systems.

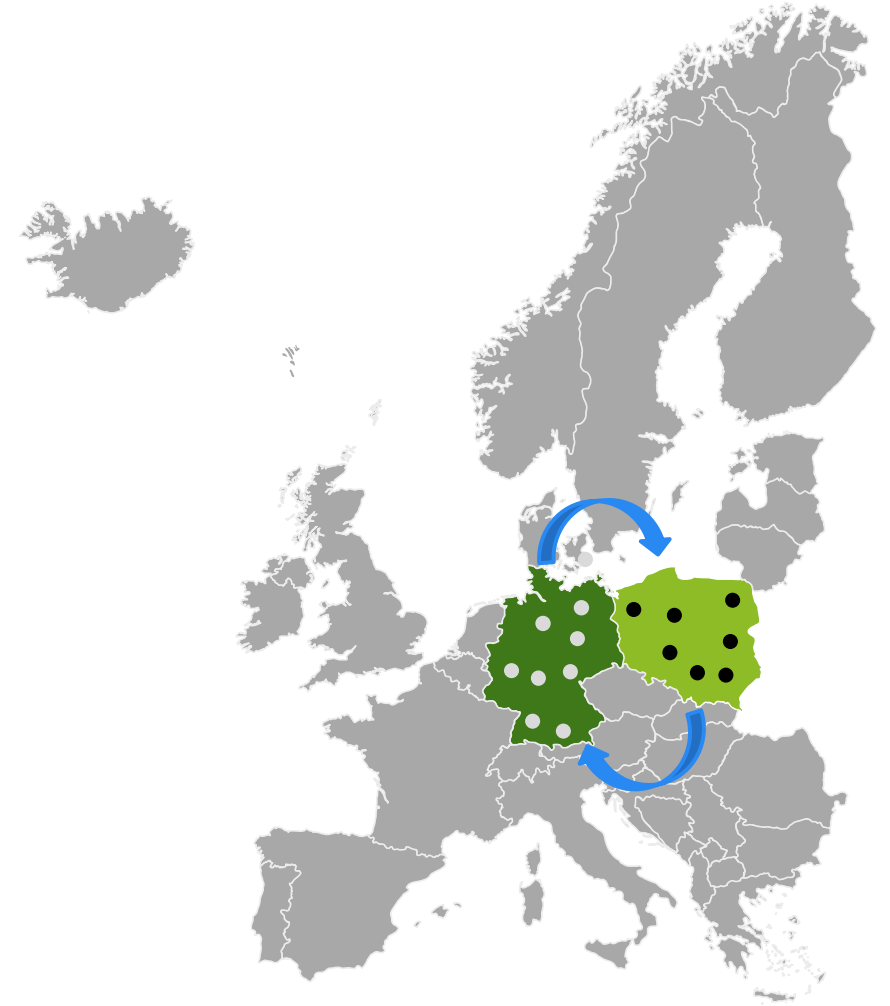
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- A joint analysis of data allows assessing the similarity in performance of those varieties tested in both countries.



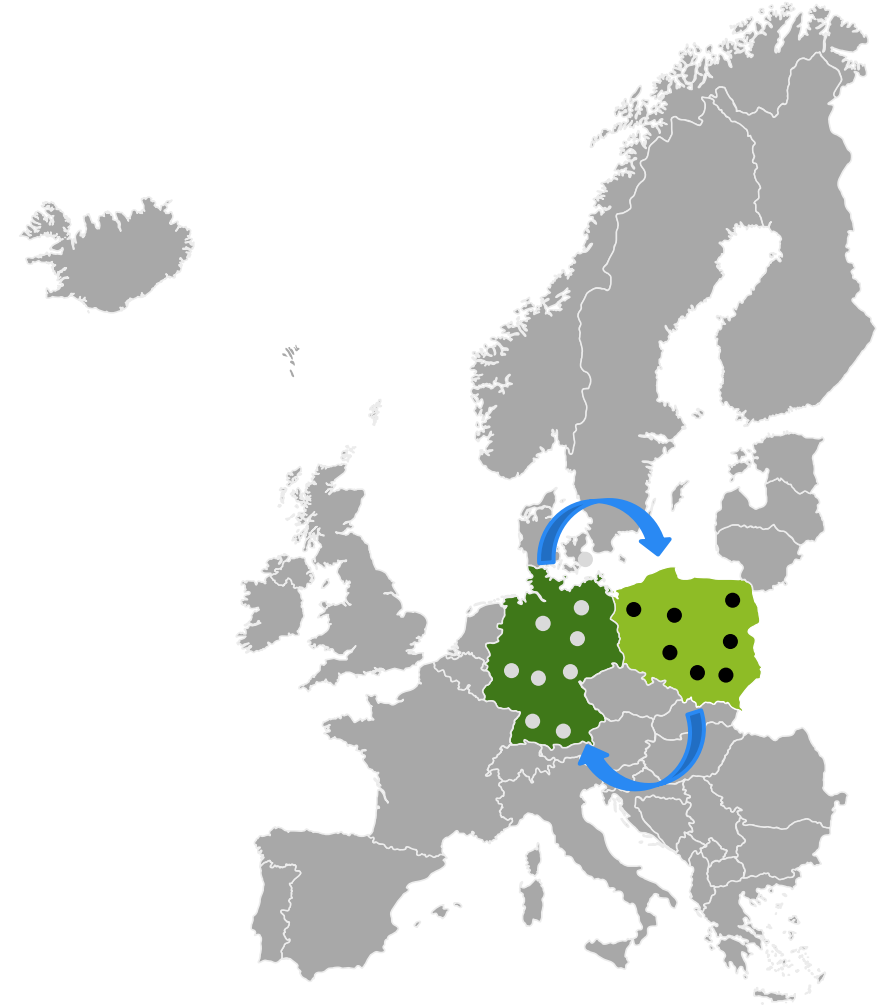
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- Varieties that are in common between countries can be regarded as checks providing some degree of connectivity between countries.
- This gives us an opportunity to borrow strength from the neighboring countries.



Joint statistical model

- A joint model for combined analysis of two or more countries data

$$yield_{ijkl} = \mu_l + g_{il} + y_{kl} + (gy)_{ikl} + s_{jl} + (sy)_{jkl} + (gs)_{ijl} + (gsy)'_{ijkl}$$

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- $(gy)_{ikl}$, $(sy)_{jkl}$, $(gs)_{ijl}$ and $(gsy)'_{ijkl}$ are random interactions.

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Assess the performance of the networks

- Using variance components from historical data
- We computed prediction error variances of a difference for BLUPs of two varieties in the same country using the plug-in approach of Stroup (2002)

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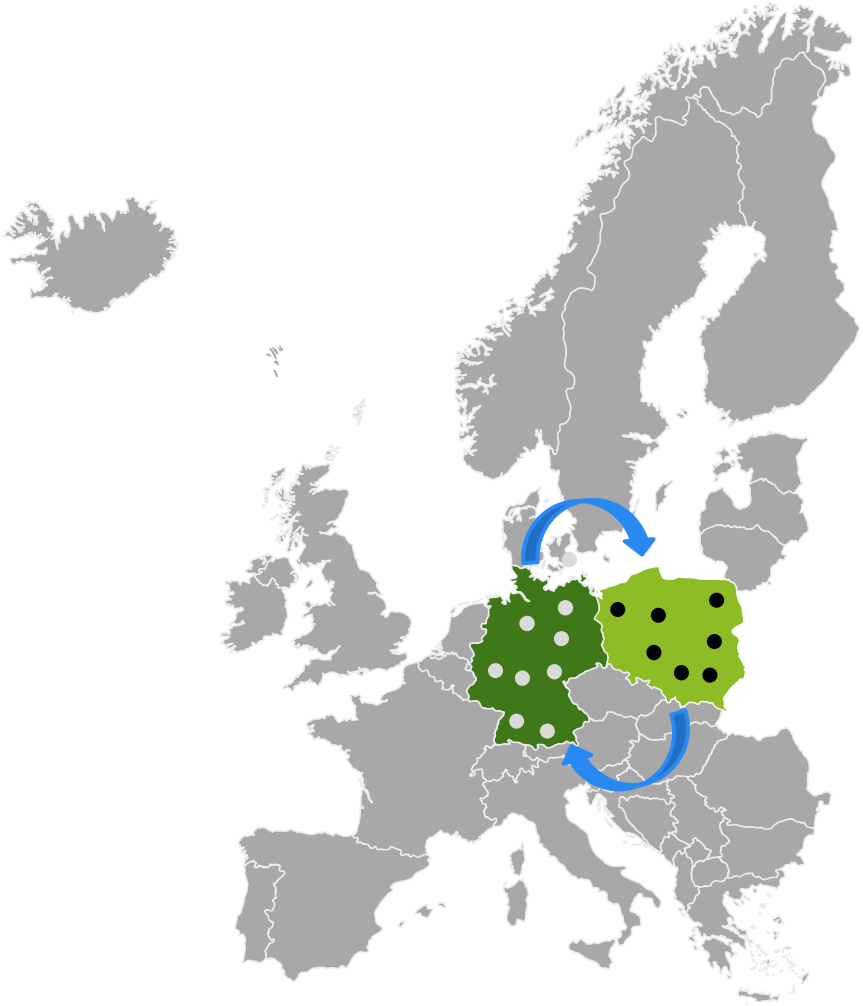
- From these variances of a difference, we computed pairwise heritabilities as given in Piepho et al (2024).

$$h_{ii'(l)}^2 = 1 - \frac{vd_{ii'(l)}}{\text{var}(g_{il} - g_{i'l})}$$

Historical maize VCU trial data in Germany and Poland

- Grain maize datasets of registered varieties from German and Polish VCU trials (Malik, 2022)
- Variety means per year and location are provided in the dataset

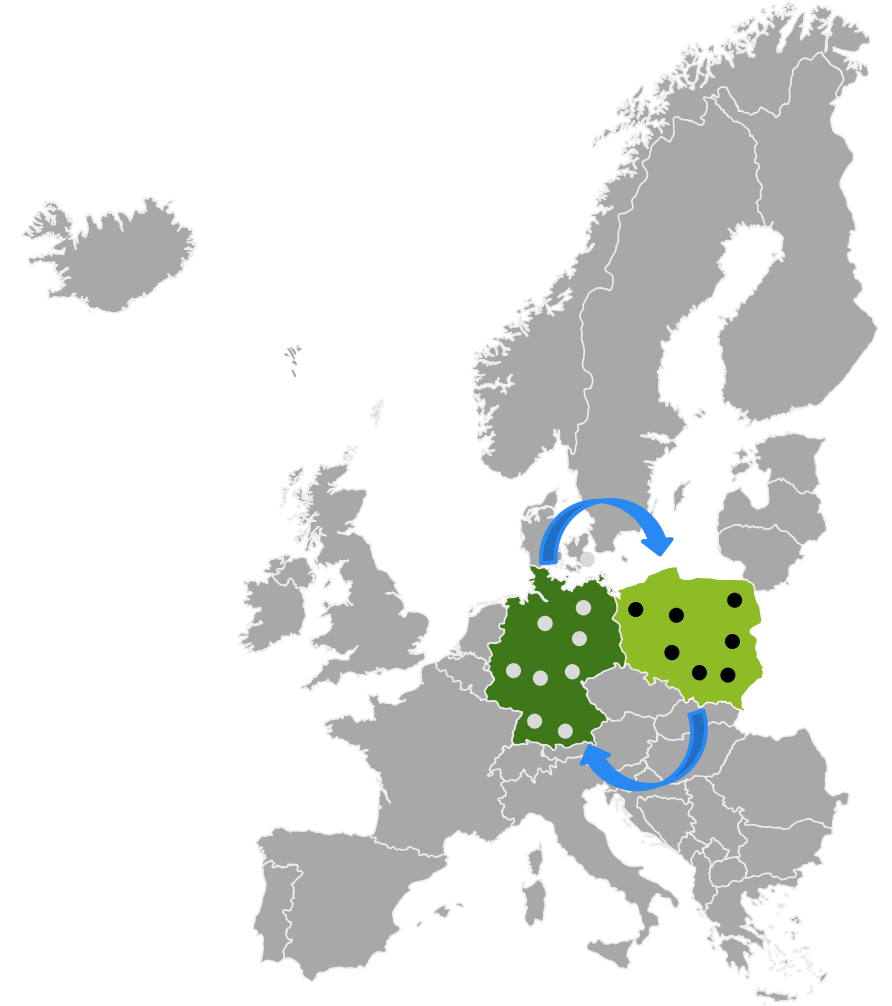
	Germany	Poland
Years	1987 – 2016	1994 – 2017
Varieties	350	634
Locations	121	32
Common varieties	43	



Joint analysis of German and Polish VCU trials

- Variance components from Malik (2022)

	Germany	Poland	$\rho_{(\text{Germany,Poland})}$
G	11.292	15.021	0.890
Y	59.706	191.284	0.812
GY	5.144	6.269	0.126
S	99.976	89.923	
GS	4.356	8.462	
YS	155.660	172.154	
Residual (GLY [~])	27.775	27.359	



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- Each trial is assumed to have a testing capacity of 200 plots.
- In total, the systems of the two countries combined require a total capacity of $200 \times 3 \times 12 \times 2 = 14,400$ plots.
- Furthermore, we assume that a total of 180 varieties are tested across the two countries.

Evaluating alternative design options

- Three MET designs across both countries will be considered.
 1. This MET design reflects the current practice with assuming 100 varieties tested in each country, from which 20 varieties are tested in both countries (180 varieties in total). For simplicity, we assume completely randomized design with two replicates.

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 3. In this MET design, we modify the individual trial design. As the capacity is 200 plots, we have 20 additional plots, on which 20 of the 180 varieties can be duplicated, leading to a partially replicated (p-rep) design.

Assessing the performance of the networks

MET design (1) (status quo): 20 varieties in common + 80 varieties in each country

Comparison type	Test country		Prediction for	Heritability ($h^2_{ii'}$)	
	Variety 1	Variety 2		Analysis scenario 1 (Correlation=0)	Analysis scenario 2 (Correlation=unstructured)
1	Germany + Poland		Germany	0.82091	0.85993
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5	Poland		Germany	0	0.65389
6	Poland		Poland	0.82552	0.82552
7	Germany	Poland	Germany	0.40636	0.73278
8	Germany	Poland	Poland	0.40864	0.82552

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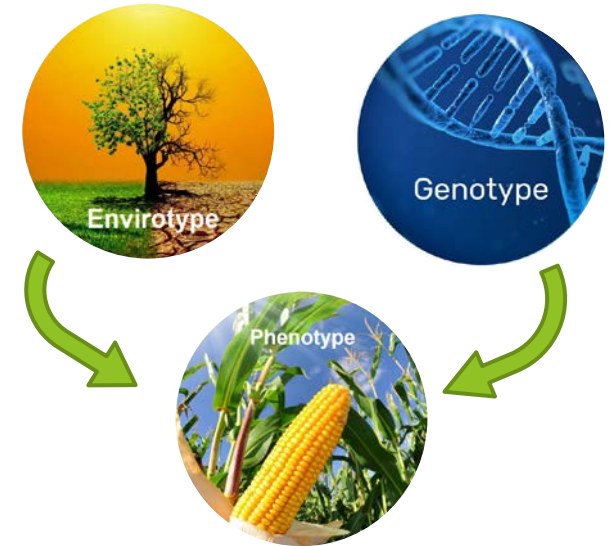
1	Germany + Poland		Germany	0.79682	0.84406
2	Germany + Poland		Poland	0.80479	0.84758

MET design (3): All 180 varieties tested in both countries, 180 varieties per trial, p-rep design per trial

1	Germany + Poland		Germany	0.83136	0.86674
2	Germany + Poland		Poland	0.83332	0.86769

Incorporating marker and environmental information

- Leveraging marker and environmental information allows for more precise estimation of genetic effects.
- Environmental information (covariates) from environmental data and soil properties.
- Fitting the same model but with regression on environmental covariates.



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- Extension of the integration to more than two countries is straightforward.
- More precision can be achieved by adding marker and environmental information.

Reference

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Thank You for your Attention !

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