

# Impact of the parameterization of soil hydrological properties on APEX model performance

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### Introduction

#### •Agriculture & Environmental Services

- •Supports crop production, pollution control, nutrient recycling, and climate regulation.
- •Soil health and land management impact environmental sustainability.

#### •Conventional Tillage (CT) Systems

- •Used in 87% of agricultural lands (Kassam et al., 2018).
- •Linked to soil degradation, erosion, and reduced water quality.

#### •Conservation Agriculture (CA)

•Based on minimal soil disturbance (no-till), permanent soil cover, and crop diversification.

•Improves soil health, organic carbon, and resilience to climate change.



### Introduction

### •Hydrological Models (Physically-Based Models)

Simulate soil, crop, and environmental dynamics.

Useful for comparing land management strategies (e.g., tillage vs. no-till).

### •Applications & Challenges

Requires extensive field data (weather, soil properties, and management practices).

Costly field experiments and limited data availability pose challenges.

Commonly used models include APEX to assess soil, water, nutrient dynamics, and climate impacts.

#### •Challenges

Requires extensive field data (weather, soil properties, management).

Field experiments are costly, and data availability is limited.

#### •APEX Model Overview

Designed to simulate small watersheds and field-scale areas.

Includes nine components: weather, hydrology, soil temperature, tillage, and more.

#### • Applications of APEX

Analyze effects of management practices on water, soil, and nutrient dynamics. Calibration and validation with field data are crucial for accuracy.





# Objective

The aim of this study is to assess the impact of soil hydraulic input parameters on simulations crop yield and model performance under conventional tillage and no-till systems.

To achieve this aim, the following research questions are stated:

- 1. Can the soil hydraulic parameters estimated with pedotransfer functions (PTFs) replace extensive field soil sampling.
- 2. Can the required data alone lead to a good model performance for both tillage systems?
- 3. How big is the variation between the model with PTFs on the simulated crop yields?



Long-Term Field Experiment

# Methodology - Experimental Site

Study Site

Long-Term Field Experiment (LTE) in Hollabrunn, Lower Austria

- Semi-arid Pannonian climate (precipitation: 493 mm/year)
- Soil: Calcareous Chernozem, loamy silt

**Tillage Treatments** 

- Conventional Tillage (CT): Moldboard plow, rotary harrow
- No-Till (NT): Direct seeder





ст	rubber	eibeneg-	ektsaat	eibeneg-	ektsaat	ст	rubber	ektsaat	eibeneg-	rubber	ст
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# Methodology - Data Collection

### **Data Collection & Model Inputs**

Soil Sampling (April 2023)

Bulk density, particle size, water retention, and hydraulic conductivity measured (e.i. Field Capacity (FC), Wilting Point (WP), Saturation (SAT))

### **APEX Model Setup**

Potential Evapotranspiration Equation: Hargreaves

Richards-based soil Water Percolation Method

10-year warm-up (1996-2023)

Input data: weather, soil, field operations (crop cycles from 2003-2023)



# Methodology – Input soil data

### Soil Hydraulic Properties & Pedotransfer Functions (PTFs)

### Soil Hydraulic Properties for Model Input

- Field Capacity (FC),
- Wilting Point (WP),
- Saturation (SAT)

### Pedotransfer Functions (PTFs)

- Used when measured values are unavailable
- Test impact of PTF-based estimations on crop yields simulation.

Variabl	e Variable
Z	Depth from the soil surface to the bottom of the layer (m)
BD	Bulk Density (t/m3)
SAN	Sand Content
SIL	Silt Content
PH	Soil pH.
WOC	Organic carbon concentration (%)

Predictor variables	FC	UW	SATC	Model
USSAND+USSILT+USCLAY+DEPTH_M	PTF01	PTF01	PTF01	PTF01.01.01
USSAND+USSILT+USCLAY+DEPTH_M+OC	PTF02	PTF02	PTF02	PTF02.02.02
USSAND+USSILT+USCLAY+DEPTH_M+BD	PTF01	PTF01	PTF01	PTF01.01.01
USSAND+USSILT+USCLAY+DEPTH_M+PH_H2O	PTF01	PTF01	PTF01	PTF01.01.01
USSAND+USSILT+USCLAY+DEPTH_M+OC+BD	PTF02	PTF02	PTF02	PTF02.02.02
USSAND+USSILT+USCLAY+DEPTH_M+OC+PH_H2O	PTF02	PTF02	PTF02	PTF02.02.02
USSAND+USSILT+USCLAY+DEPTH_M+BD+PH_H2O	PTF03	PTF01	PTF05	PTF03.05.01
USSAND+USSILT+USCLAY+DEPTH_M+OC+BD+PH_H2O	PTF02	PTF07	PTF02	PTF02.02.07



# Methodology - Experimental Site

### Calibration & Validation Process Model Calibration

- Calibrated yields on Block A
- Sensitivity analysis to identify key parameters for yield
- Calibration performed on yield data for maize and winter wheat

Validation

- Yields on blocks B and C to validate
- Evaluating the influence of PTFs on model accuracy

Parameter	Definition				
PARM23	Hargreaves PET equation coefficient				
PARM34	Hargreaves PET equation exponent				
PARM26	Fraction of maturity at spring growth initiation				
WA	Biomass-Energy Ratio				
ні	Harvest Index				
ТОР	Optimal Temperature for Plant Growth				
TBS	Minimum Temperature for Plant Growth				
DMLA	Maximum Potential Leaf Area Index				
	Fraction of Growing Season When Leaf Area				
	Declines				





# Methodology - Analysis of Interaction Effects: PTF and Tillage System on Crop Yields

Examine the interaction between soil hydraulic properties and tillage systems on yields of CORN and WWHT.

### Method:

Utilized a generalized linear mixed-effects model (GLMM) to evaluate the relationship between yields (response) and predictors (soil properties & tillage).

Random intercepts for tillage systems were included to account for treatment variation.

### Model:

```
Yields ~ Model*(1 | Treatment)
```

### **Robust Estimation:**

Used the robustImm R package for model fitting.

Ensures reliable estimates in the presence of potential data contamination by applying random effects contamination models.



### Results: APEX Calibration and Performance

The initial setup lead to 21 optimal, non-unique sets of crop parameters were identified across various crops and tillage systems.

Using a shared set of crop parameters for all treatments led to inconsistent performance, particularly in validation.

Reduced accuracy for CORN yields under non-tillage systems highlights sensitivity to soil and crop input conditions.

Table 4: Model performance statistics for crop yields under different treatments and blocks.

Gran	Troatmont	Plock <sup>1</sup>	RSI	ME	KGE		
Стор	freatment	DIOCK	Min	Max	Min	Max	
	IST	А	1.04	1.12	0.50	0.58	
	NTS	А	0.85	0.97	0.45	0.57	
	IST	В	0.97	1.05	0.49	0.53	
VV VV H I	NTS	В	0.79	0.92	0.50	0.64	
	IST	С	0.81	0.92	0.53	0.64	
	NTS	С	0.95	1.07	0.46	0.59	
	IST	А	1.05	1.17	0.63	0.70	
	NTS	А	0.93	1.10	0.49	0.55	
CORN	IST	В	0.83	1.12	0.39	0.82	
CORN	NTS	В	0.91	1.37	-0.55	-0.33	
	IST	С	0.74	1.19	0.61	0.78	
	NTS	С	1.03	1.37	0.07	0.17	



Figure 1: Model yield versus historical LTE-reported (2006-2023) yield for WWHT and CORN under ITS and NTS tillage system.



### Results: APEX Calibration and Performance

The best-fit solutions for non-unique crop parameter sets resulted in the following outcomes:

- In the ITS, the simulated average yields for corn and wheat were 10.1 Mg/ha and 6.7 Mg/ha, respectively, with root mean square errors (RMSE) of 1.04 and 1.05.
- In the NTS, the average yields for corn and wheat were 10.8 Mg/ha and 6.2 Mg/ha, respectively, with RMSE values of 0.85 and 0.93.



Figure 1: Model yield versus historical LTE-reported (2006-2023) yield for WWHT and CORN under ITS and NTS tillage system.



## Results: APEX performance with PTFs



To evaluate the impact of PTFs model parameters on Yields simulation, values for the RSME and KGE performance criteria were compared between the 21 optimal non-unique sets of crop parameters selected across different crops and tillage systems.

Model setup with soil data derive from an extensive soil sampling poorly performed for CORN under non-till system.

**Figure 2:** Comparison of performance criteria between model simulations. The horizontal red slash line shows optimal value of performance criteria.



### Results: Estat. effects of PTFs and tillage on Yields

### **Effects on WWHT Yields**

- Baseline Intercept:
- Measured Data: Decreased yields by 0.752.
- PTF Models:
  - PTF02.02.02: Decreased yields by 0.267.
  - PTF02.07.02: Decreased yields by 0.288.
  - PTF03.01.05: Decreased yields by 0.209.
  - PTF01.01.01: Increased yields by 0.286.
- **Tillage System**: No significant effect on WWHT yields.



*Figure 2: Model yield versus historical LTE-reported (2006-2023) yield for WWHT and CORN under ITS and NTS tillage system.* 



# Effects of PTFs and tillage on Yields

### **Effects on CORN Yields**

- Baseline Intercept: 11.2 Mg/ha .
- Measured Data: Decreased yields by 0.8 Mg/ha
- PTF Models:
  - PTF01.01.01: Increased yields by 0.089.
  - PTF02.02.02: Increased yields by 0.029.
  - PTF02.07.02: Increased yields by 0.027.
  - PTF03.01.05: Negligible effect.
- **Tillage System**: Weak sig. impact (variance: 0.004925).





Figure 2: Model yield versus historical LTE-reported (2006-2023) yield for WWHT and CORN under ITS and NTS tillage system.



### Conclusion

### **PTFs vs. Measured Values**

Soil hydraulic properties significantly impact APEX simulations.

PTFs can introduce uncertainty, particularly in no-till (NT) systems (KAST).

### **Key Findings**

Accurate soil hydraulic data enhances model performance, improving yield and environmental predictions.

PTFs are useful but can increase uncertainty, especially in NT systems.

Proper parameterization is crucial for reliable simulations.

### On going Research

Investigate the impact of extreme climate scenarios on tillage systems and model accuracy.



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https://ejpsoil.eu/soil-research/eom4soil/into-dialogue/soilx

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