

Long-term no-tillage farming mitigates waterlogging during extreme rainfall events in black soil (Mollisols) region of Northeast China

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ABSTRACT

The increasing frequency of extreme rainfall events intensifies waterlogging-induced aeration stress on crops, highlighting the urgency for adaptive agricultural practices. This study evaluates the potential of long-term no-tillage (NT) farming to mitigate waterlogging risks on a silty clay loam in Northeast China by comparing with traditional moldboard plough (MP) farming using an 11-year field experiment. Rainfall simulation (40 mm h⁻¹) and dye-tracing were employed to study the dynamics water infiltration, and X-ray computed tomography (CT) was applied to examine soil pore characteristics (0–60 cm depth) and their responses (0–5 cm layer) to raindrop splashing. Results demonstrated that NT significantly enhanced water infiltration capacity, with a 161 % higher maximum infiltration depth (*MID*), an 65 % increase in total stained area (*TSA*), and 95 % less surface ponding compared to MP. Although MP exhibited higher total porosity (ϵ_{total}), macroporosity (>0.04 mm pores, $\epsilon_{\text{X-ray}}$), and increased pore connectivity in the 0–20 cm layer, it had fewer biopores and greater vertical stratification of macroporosity compared to NT across the 0–60 cm profile. During rainfall events, the MP plot experienced a 28 % reduction in porosity of pores sized 0.2–1.0 mm and a 19 % increase in porosity of pores <0.2 mm in the surface layer, likely due to raindrop splashing, while NT maintained pore integrity through straw mulching. We conclude that long-term NT is effective in enhance soil resilience to waterlogging by developing continuous vertical pore networks and maintaining surface mulch, thereby improving climate adaptability in the black soil region of Northeast China.

1. Introduction

Global climate change has intensified extreme rainfall events, characterized by heavy rainfall or prolonged droughts (Fischer and Knutti, 2015; Piao et al., 2010). Heavy rainfall can trigger overland runoff, soil erosion, and crop hypoxia stress, thereby threatening global food security (Hess et al., 2018). Soil water infiltration capability is a primary defense against heavy rainfall, as it directly determines the partitioning of rainfall at the soil surface. A high infiltration capacity allows more water to enter the soil profile, while a low capacity causes water to accumulate on the surface or in the root zone during or after downpour events. This capability is strongly influenced by farming practices

(Cynthia et al., 2002; Hess et al., 2018). Conservation management like no-tillage (NT) farming offers agronomic and environmental benefits and are widely adopted (Fernández-Ugalde et al., 2009; Govindasamy et al., 2020). However, there lacks consensus on the impact of NT on water infiltration capacity (e.g., Blanco-Canqui et al., 2017; Fan et al., 2013; Moraes et al., 2016). This disagreement may stem from how NT management shapes the soil pore network and how this interacts with rainfall conditions.

A key mechanism through which tillage practices influence infiltration is by altering soil pore architecture. Previous studies have primarily assessed water infiltration into soils using the methods such as constant head infiltration on intact cores in lab, single or double-ring infiltration

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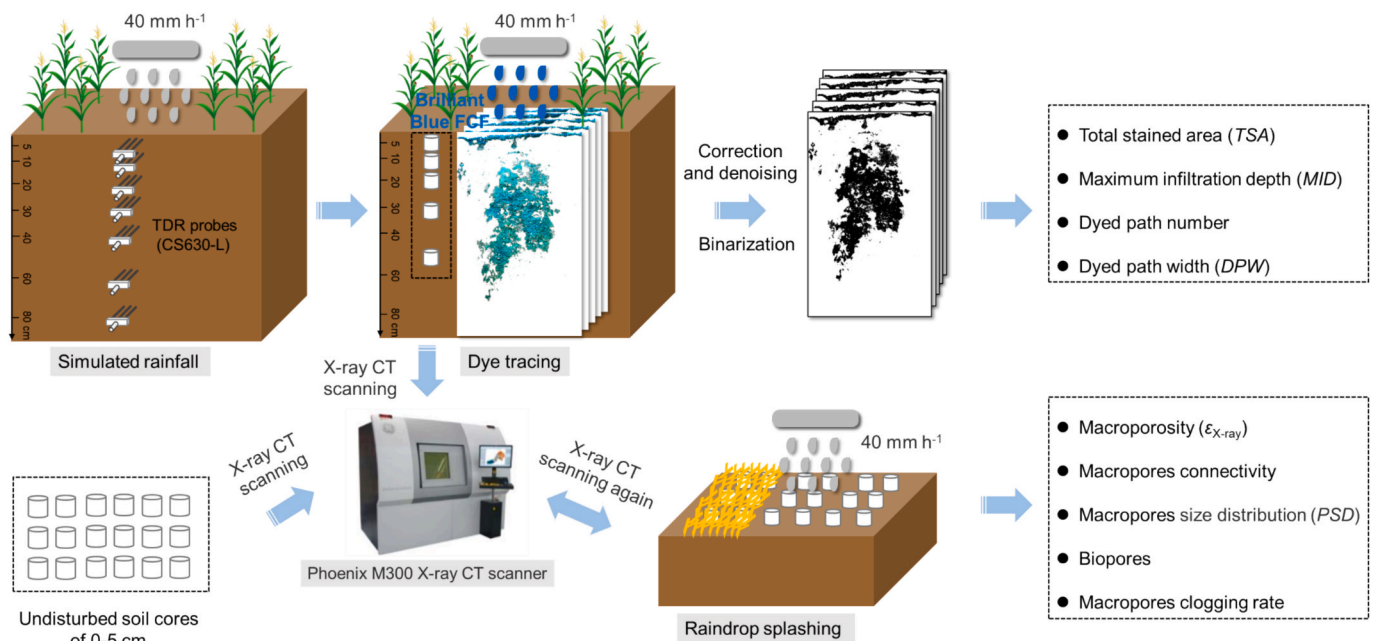


Fig. 1. Schematic diagram of experimental design.

in field, dye tracing, and various infiltration models (Fan et al., 2013; Haruna et al., 2024; Soracco et al., 2019). Specifically, the co-occurrence and connectivity of macropores throughout the soil profile are crucial for facilitating rapid water movement via preferential flow paths (Allaire-Leung et al., 2000). However, soil core results inadequately represent profile-scale infiltration because they sample a limited volume and miss structural connectivity (e.g., macropore connection) between soil layers, an important factor influencing soil water movement. Moldboard plow (MP) farming typically increased topsoil porosity and connectivity but can disrupt subsoil pores and created a plow pan, reducing vertical connectivity due to tillage disturbance (Wardak et al., 2024). In contrast, long-term NT management promotes the development of stable biopores (e.g., former root channels and earthworm burrows) and enhances their continuity across soil layers, creating a more connected and efficient pore network (Budhathoki et al., 2022; Wardak et al., 2024). These biopores act as hydraulically efficient preferential flow paths, transferring water more rapidly and deeply into the subsoil than irregular non-biopores due to their greater structural stability, lower tortuosity, and smoother walls (Mestelan et al., 2021; Pagenkemper et al., 2015; Zhang et al., 2019). However, NT can also lead to higher surface bulk density, particularly in fine-textured soils, which may initially hinder infiltration (Blanco-Canqui et al., 2017), highlighting the complex trade-off between pore connectivity and topsoil conditions under NT.

Although field measurements using single or double-ring infiltrometers and dye tracing can reflect the effects of pore networks within soil profile, they are often conducted under simulated ponding without rainfall, thus failing to account for the disruptive effects of raindrop impact on surface soil structure. Raindrops dynamically alter soil structure during rainfall, with splashing disaggregating soil particles and sealing surface soil pores, potentially reducing infiltration by 11–56 % (Assouline, 2004; Legout et al., 2005; Ma et al., 2022). A key advantage of NT systems is the presence of a protective straw mulch buffer layer, which can effectively mitigate raindrop splash's negative effects on soil structure under NT (Almeida et al., 2018; Gómez-Paccard et al., 2015). Consequently, infiltration measurements under non-rainfall conditions may misrepresent the actual effects of tillage management, particularly for NT whose benefits are tied to both subsurface pore connectivity and surface protection.

In this study, we hypothesized that long-term NT farming enhances

infiltration during extreme rainfall by maintaining profile-wide macropore continuity and straw mulching's buffering effects. The objectives of this study were to (1) characterize water infiltration and flow patterns under NT and MP during extreme rainfall events, and (2) explore the underlying mechanisms from the perspective of soil pore system—considering both surface protection and subsurface connectivity—using high-frequency soil water content monitoring, dye tracers, and X-ray CT.

2. Materials and methods

2.1. Study site

We used a long-term farming system experiment initiated in the spring of 2011 at the Lishu Experiment Station of China Agricultural University, Jilin Province (43°16' N, 124°26' E), China. The region has a sub-humid monsoon climate with an average annual temperature of 5.9 °C and precipitation of 556 mm. The soil is classified as Mollisol (Mollisol, USDA) with a silty clay loam texture (17.2 % sand, 46.1 % silt, and 36.7 % clay) and an organic matter content of 21.0 g kg⁻¹ (Soil Survey Staff, 2022).*

This study compared two treatments: no-tillage (NT) farming and traditional moldboard plough (MP) farming, both with mono-cropped maize (*Zea mays* L.). Each treatment had three replications, with plots measuring 21.6 m by 63.0 m. Under NT, ~6000 kg ha⁻¹ of crop residue remained on the soil surface at ~30 cm stubble height. For the MP treatment, maize residue was removed from the field after harvesting and soil was tilled to a depth of about 20 cm using a moldboard plow the following spring. For both treatments, maize was planted in May using a paired-row no-till seeder (2BMZF-2, Jilin Province Kangda Agricultural Machinery Group Co. Ltd.). The two treatments received equal fertilizer application (800 kg ha⁻², N-P₂O₅-K₂O (26-10-12)) and were maintained at a planting density of 63,000 seeds per hectare. No irrigation was applied during the growing season. We applied herbicides at post-sowing to control weeds in both treatments. Maize was harvested using a combine harvester (4YL-6(R230), John Deere (Jiamusi) Agricultural Machinery Co. Ltd.) in the fall.

2.2. Field studies on soil water infiltration

2.2.1. Simulated rainfall experiments

Simulated rainfall experiments were conducted to compare the infiltration characteristics of the NT and MP systems. The rainfall-simulator consisted of a peristaltic pump (FB-I/104kA, Chongqing Jieheng Peristaltic Pump Co. Ltd.) that maintained a 4 cm water column pressure between two 50×50 cm acrylic glass plates, each equipped with 196 evenly spaced 26 G cannulas (Marquart et al., 2020) (Fig. S1). The rainfall-simulator was mounted on a 70 cm-high steel frame, positioned over a 50×50 cm plot, and enclosed with plastic film to prevent wind interference with raindrop trajectories. To minimize lateral water seepage, a stainless-steel frame (10 cm in height, 0.2 cm thickness) was inserted 5 cm into the soil beneath the setup. A 10-cm long steel ruler was placed vertically within the plot to measure surface water depth at a 10-minute interval. The rainfall intensity was set to 40 mm h^{-1} based on local extreme rainfall data (Fig. S2). Each tillage treatment underwent six replicated rainfall simulations, with each event lasting one hour. Adjacent events were spaced several hours apart to ensure consistent initial volumetric water content across replicates (Table S1).

During the rainfall simulation studies, volumetric soil water content (θ) was monitored using time-domain reflectometry (TDR). A pit was excavated at the infiltration site, and seven pre-calibrated TDR probes (CS630-L, Campbell Scientific, Logan, UT, USA), with 15 cm-long, 0.32 cm-diameter rods were installed at depths of 5, 10, 20, 30, 40, 60, and 80 cm in both NT and MP plots (Fig. 1). To reduce soil disturbance, the probes were installed 45 days before the rainfall simulation experiment. The sensors were connected to a TDR200 system (Campbell Scientific, Logan, UT, USA) controlled by a CR1000X data logger (Campbell Scientific, Logan, UT, USA). The θ values were recorded at a 3-minute interval to accurately capture infiltration dynamics.

2.2.2. Dye tracer experiment

To visualize water flow paths under NT and MP treatments, we performed a dye-tracing experiment immediately after the simulated rainfall study in the same plots. We used Brilliant Blue FCF (4 g L^{-1}) as the tracer due to its low toxicity, high visibility, and excellent mobility (German-Heins and Flury, 2000; Marquart et al., 2020). The rainfall intensity and duration matched those of the previous experiment. Forty-eight hours after dye application, we excavated five soil pits (75 cm deep and 45 cm wide) to examine dye distribution through the soil profile. Soil profiles were photographed using a Sony $\alpha 6400$ camera (4912×3264 pixels, Fig. 1). We employed a linen cloth during photography to maintain consistent lighting and eliminate direct sunlight interference.

2.3. Characteristics of soil pores

For soil analysis following the rainfall simulation and dye-tracer experiments, we collected undisturbed soil cores from five depth intervals (0–5, 5–10, 10–20, 20–40, and 40–60 cm) using PVC tubes (5.0 cm height \times 4.8 cm inner diameter, Fig. 1). Five replicate cores were randomly sampled from each layer. The cores were scanned using a Phoenix M300 X-ray CT scanner (GE Sensing and Inspection Technologies, GmbH, Wunstorf, Germany) at 160 kV beam energy and 150 μA current. After scanning, samples were oven-dried at 105°C for 24 h, cooled to room temperature, and weighed to determine bulk density. Then, the total porosity (ϵ_{total}) was calculated with a particle density of 2.65 g cm^{-3} .

2.4. Raindrop splashing

To quantify splash erosion on the surface soil structure during simulated rainfall (Fig. 1 and S3), we conducted a raindrop splash experiment using 18 undisturbed soil cores (0–5 cm) collected near the experimental area in PVC tubes (5.0 cm height \times 4.8 cm inner diameter). The samples included 12 cores were from NT plots (6 with maize straw

mulch [NT], 6 bare [NT-S]) and 6 cores were from MP plots (Fig. S4). All cores were initially CT-scanned, then exposed in situ to 40 mm h^{-1} rainfall (match the simulation protocol). After splash exposure, samples were carefully retrieved and re-scanned using identical CT parameters (160 kV and 150 μA).

2.5. Image analysis

For the dye tracer experiment, all images were geometrically corrected and denoised using Adobe Photoshop CC2018 (Adobe, San Jose, CA, USA), resulting in processed images measuring 75×45 cm with 3750×2250 pixels resolution. Following binarization in Image J (Version 1.50i, National Institute of Health, <http://imagej.nih.gov/ij/>), we classified soil profile into dyed and non-dyed regions.

Using MATLAB R2022a (MathWorks, Natick, MA, USA), we quantified four infiltration parameters: total stained area (TSA), maximum infiltration depth (MID), number of dyed paths, and dyed path width (DPW) as a function of depth (Fig. 1). DPW variations revealed three distinct water flow patterns: (1) <20 mm: preferential flow with minimal matrix interaction; (2) 20–200 mm: preferential flow with moderate matrix interaction or non-homogeneous matrix flow; (3) >200 mm: homogeneous matrix flow (Weiler and Flühler, 2004; Zwartendijk et al., 2017).

For the X-ray CT analysis, each soil core was scanned to acquire 1500 projections, which were reconstructed into 3D tomograms using Datas \times 2.0 software (GE Sensing and Inspection Technologies, GmbH, Wunstorf, Germany) (Fig. 1). The reconstructed 8-bit greyscale images (40 μm resolution) were processed in ImageJ (Version 1.50i, National Institute of Health, <http://imagej.nih.gov/ij/>) with the following protocols. For general analysis, (1) A cylindrical region of interest (ROI, 920×789 pixels) was selected from the core center; (2) A Gaussian filter (radius = 1) was applied to reduce image noise; and (3) Pores were segmented using default thresholding (Yudina et al., 2022). For the raindrop splash analysis, an extended ROI (1100×1152 pixels; 0.1–4.7 cm depth range) captured surface soil structure changes, and identical filtering and segmentation methods were applied.

Macroporosity ($\epsilon_{\text{X-ray}}$) and macropore connectivity were quantified at 40 μm resolution. Connected macropores were identified using Avizo 9.0's 'Axis Connectivity' module (Thermo Fisher Scientific, Waltham, MA, USA) (Lu et al., 2019), with connectivity expressed as the volume ratio of connected macropores to $\epsilon_{\text{X-ray}}$. Macropores size distribution (PSD) was determined via the maximum inscribed spheres method, using the Thickness plugin in Image J (Hildebrand and Rüeggsegger, 1997; Houston et al., 2017). Biopores were identified at 40 μm resolution using Fiji's 'Biopores' plugin (Version 1.53t, National Institute of Health, <https://imagej.nih.gov/ij/>), followed by removal of misclassified features using Avizo 9.0's 'Filters' module (Lucas et al., 2019).

2.6. Data calculation and statistical analysis

2.6.1. Data calculation

To evaluate water infiltration during the simulated rainfall period, we first determined the maximum infiltration depth (MID, cm) using the two-curve methods. This approach identifies the crossover points between soil water content profiles measured before and after rainfall (Guo, 2020). Based on the MID, soil water supply (S_t , mm) was calculated as the difference in θ over the soil depth L (cm) before and after rainfall event:

$$S_t = \int_0^L \theta(x, t) dx - \int_0^L \theta(x, 0) dx \quad (1)$$

where t (min) is the time when θ ($\text{cm}^3 \text{ cm}^{-3}$) at depth L stops increasing during the rainfall event.

Wetting front velocity (V_{wf} , cm min^{-1}) (Hu et al., 2019) and average

Table 1

Ponding depth at different times, maximum infiltration depth (*MID*) and soil water supply (*S_i*) under no-tillage (NT) and moldboard plough (MP). Different lower-case letters indicate significant differences between tillage treatments ($P < 0.05$).

Tillage treatments	Ponding depth (cm)						<i>MID</i> (cm)	<i>S_i</i> (mm)
	10 min	20 min	30 min	40 min	50 min	60 min		
MP	0.92 ± 0.32	1.37 ± 0.25a	1.67 ± 0.31a	1.85 ± 0.38a	2.1 ± 0.53a	2.24 ± 0.61a	40	6.74 ± 2.10b
NT	0	0.01 ± 0.03b	0.04 ± 0.01b	0.06 ± 0.02b	0.06 ± 0.02b	0.12 ± 0.19b	60	13.67 ± 1.65a

wetting rate (S_{mean} , $10^{-2} \text{ cm}^3 \text{ cm}^{-3} \text{ min}^{-1}$, Tian et al., 2019) were calculated as:

$$V_{wf} = \frac{h}{t_1} \quad (2)$$

$$S_{\text{mean}} = \frac{\theta_{t_2} - \theta_{t_1}}{t_2 - t_1} \times 100 \quad (3)$$

where h (cm) is the soil thickness at various depths (5, 10, 20, 30, 40, 60, and 80 cm), t_1 (min) is the time when the probe first detects an increase in θ ($\geq 0.005 \text{ cm}^3 \text{ cm}^{-3}$, considering the measurement accuracy of the pre-calibrated TDR sensors), t_2 (min) is the time when the increase ceases, Parameters θ_{t_1} ($\text{cm}^3 \text{ cm}^{-3}$) and θ_{t_2} ($\text{cm}^3 \text{ cm}^{-3}$) represent the volumetric soil water content at t_1 and t_2 , respectively, and 100 is a conversion coefficient.

The stratification ratio (*SR*) of $\varepsilon_{X\text{-ray}}$ and bioporosity was calculated to evaluate vertical variations of pore characteristics:

$$SR_{\varepsilon_{X\text{-ray}} \text{ or bioporosity}} = \frac{(\varepsilon_{X\text{-ray}} \text{ or bioporosity})_{0-20\text{cm}}}{(\varepsilon_{X\text{-ray}} \text{ or bioporosity})_{20-60\text{cm}}} \quad (4)$$

where $(\varepsilon_{X\text{-ray}} \text{ or bioporosity})_{0-20\text{cm}}$ and $(\varepsilon_{X\text{-ray}} \text{ or bioporosity})_{20-60\text{cm}}$ are detectable macroporosity and bioporosity at the 0–20 and 20–60 cm soil layers, respectively, Higher *SR* values deviating from 1 indicate greater vertical stratification of $\varepsilon_{X\text{-ray}}$ or bioporosity across the soil profile.

To quantify structural changes caused by raindrop splash, macropores clogging rate was calculated as (Ma et al., 2022):

$$\text{Macropores clogging rate} = \frac{\varepsilon_{X\text{-ray}0} - \varepsilon_{X\text{-ray}1}}{\varepsilon_{X\text{-ray}0}} \times 100\% \quad (5)$$

where $\varepsilon_{X\text{-ray}0}$ and $\varepsilon_{X\text{-ray}1}$ are the detectable macroporosity before and after raindrop splashing, respectively.

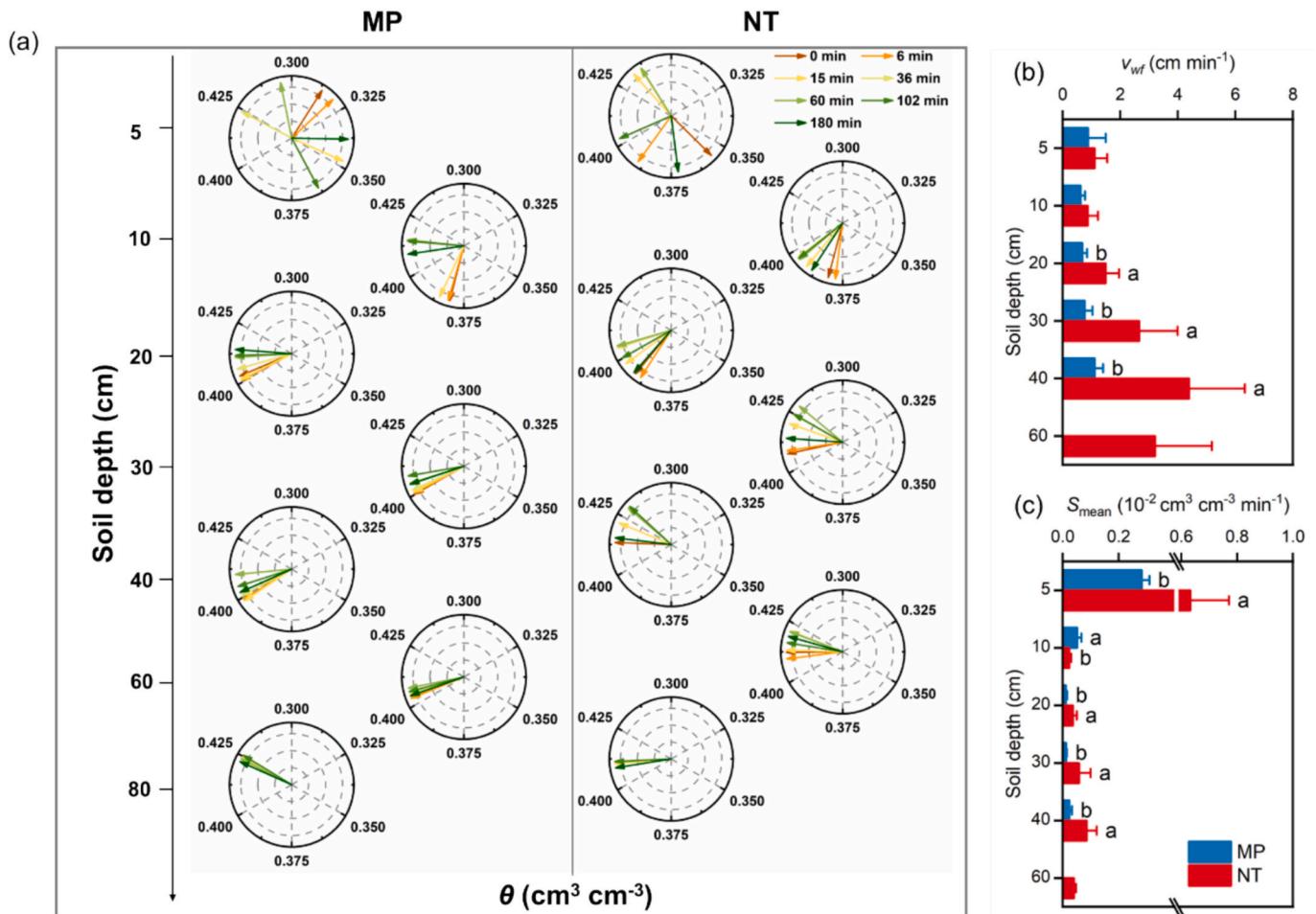


Fig. 2. The dynamic of soil water content (θ) (a), penetration velocity of the wetting front (V_{wf}) (b), and average wetting rate (S_{mean}) (c) with depth after infiltration experiment under no-tillage (NT) and moldboard plough (MP). The data is the average value ($n = 6$). Different lower-case letters indicate significant differences between tillage treatments ($P < 0.05$).

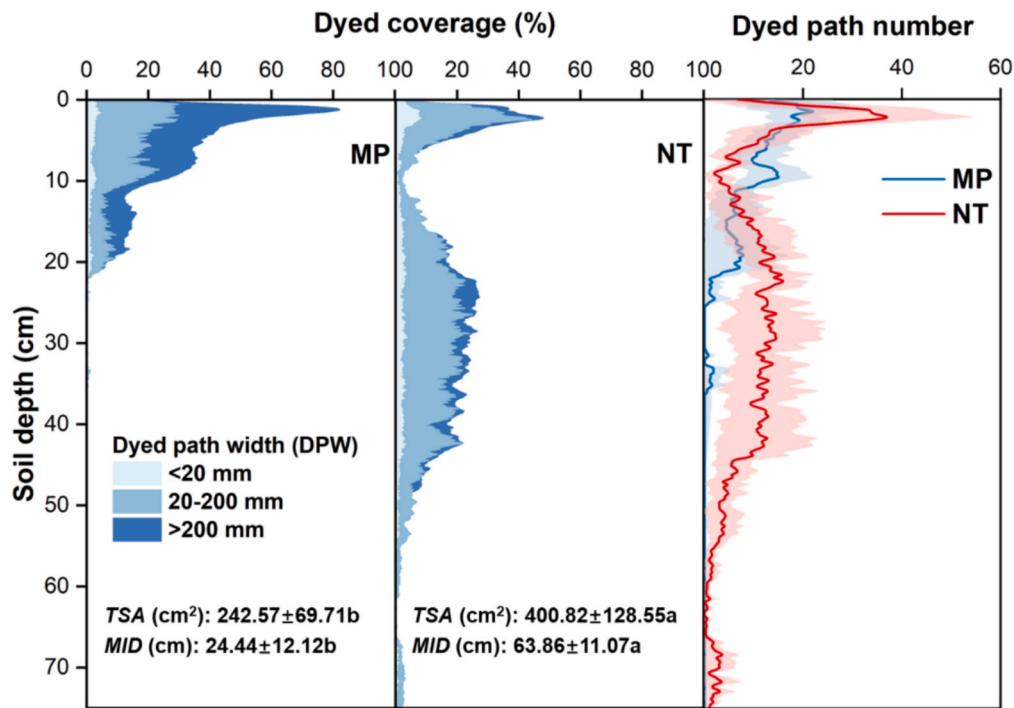


Fig. 3. The total stained area (TSA), maximum infiltration depth (MID), dyed coverage for three dyed path width (DPW) (<20, 20–200, and >200 mm), and dyed path number with soil depth under no-tillage (NT) and moldboard plough (MP). The data is the average value (n = 6). The shaded area is the standard deviation. Different lower-case letters indicate significant differences in TSA or MID between tillage treatments ($P < 0.05$).

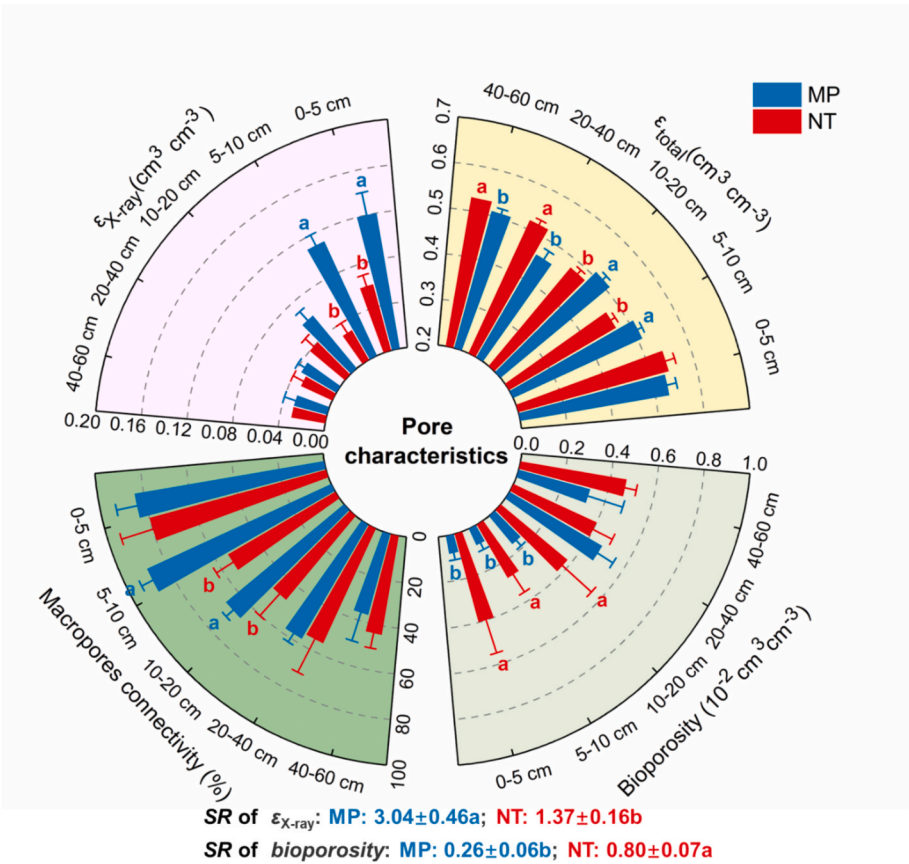


Fig. 4. Pore characteristics in the 0–60 cm soil profile under no-tillage (NT) and moldboard plough (MP). ϵ_{total} is the total porosity. ϵ_{X-ray} is the macroporosity derived from X-ray CT. SR is the stratification ratio of ϵ_{X-ray} and bioporosity (0–20 cm/20–60 cm). The data is the average value (n = 5). Different lower-case letters indicate significant differences between tillage treatments at the same soil depth ($P < 0.05$).

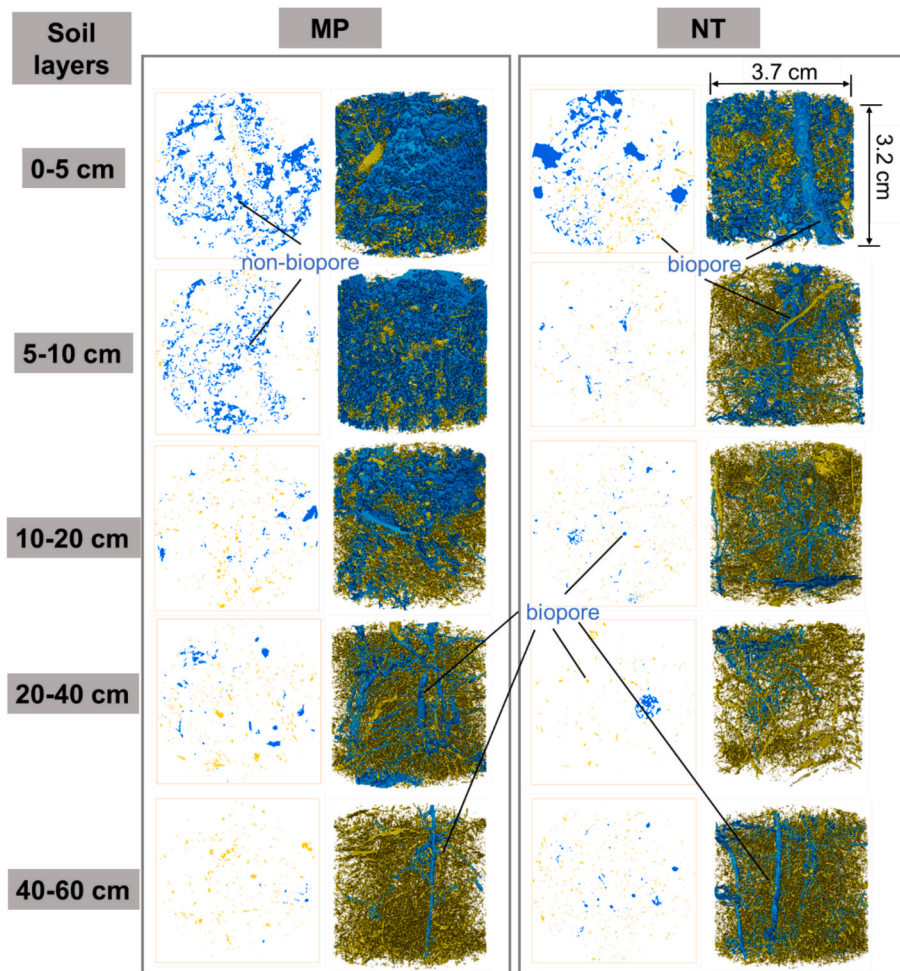


Fig. 5. 3D visualizations of soil cores of no-tillage (NT) and moldboard plough (MP) in the 0–5, 5–10, 10–20, 20–40, and 40–60 cm soil layers, respectively. The blue color represents connected pores from top to bottom of the core, and the yellow color represents isolated pores.

2.6.2. Statistical analysis

Statistical analyses were performed to compare tillage treatment effects on hydrological and pore characteristics. Student's *t*-test assessed differences in ponding depth, *MID*, S_b , V_{wf} , S_{mean} , *TSA*, and pore parameters after verifying data normality through residual plot inspection. We conducted one-way ANOVA to compare pore characteristics and macropore clogging rates across treatments. All analyses were performed with SPSS Statistics 24.0 (IBM, Armonk, NYC, USA).

We employed structural equation modeling (SEM) using IBM SPSS AMOS 24.0 (IBM, Armonk, NYC, USA) to examine causal relationships among variables. The models were developed through an iterative process of construction, validation, and adjustment. Using maximum likelihood estimation, we derived standardized path coefficients to quantify both the magnitude and direction of variable relationships.

3. Results

3.1. Characteristics of water infiltration under contrasting tillage practices

Water infiltration behavior differed markedly between the NT and MP systems. During one hour of simulated rainfall, NT maintained minimal surface ponding (0.12 cm). In contrast, MP developed substantial ponding (0.92 cm) within just 10 min, reaching 2.24 cm after one hour— a value 18 times higher than under NT ($P < 0.05$) (Table 1). Cumulative infiltration depths also varied markedly: while both treatments reached 5 cm after 6 min, NT showed substantially deeper infiltration thereafter, reaching >60 cm after 60 min, comparing to only 40

cm under MP (Fig. 2a). The *MID* under MP (40 cm) was notably lower than under NT (60 cm), and the S_b values further confirmed this trend (6.74 mm for MP vs. 13.67 mm for NT, $P < 0.05$) (Table 1). Additionally, V_{wf} at depths at the 20, 30, and 40 cm depths under MP was significantly lower ($P < 0.05$) than under NT, while S_{mean} was lower at most depths except 10 cm, where MP showed higher saturation ($P < 0.05$) (Fig. 2b, c).

Dye tracer experiments provided further insight into the mechanisms driving these infiltration patterns. The *TSA* and *MID* under NT were 1.7 and 2.6 times greater, respectively, than under MP ($P < 0.05$) (Fig. 3), indicating more extensive water movement in NT soils. Under MP, dyed coverage and the number of stained pathways decreased steadily with depth. In contrast, NT exhibited dynamic flow pathways: dye coverage initially increased (0–5 cm), declined (5–10 cm), and then expanded again at greater depths, suggesting preferential flow through biopores or root channels. Notably, MP had 32.8 % dye coverage in the 0–15 cm layer, more than double that of NT (14.3 %), but this trend reversed in deeper layers (Fig. 3). Dye pathway width also differed: MP predominantly featured wider paths (20–200 and >200 mm), while NT had a higher proportion intermediate-width path (20–200 mm). Pathway counts were higher under NT in the 0–5 cm and deeper layers but lower in the 5–15 cm layer compared to MP. These results highlight that NT promotes more heterogeneous and deeper water penetration, whereas MP restricts flow to narrower, shallower pathways.

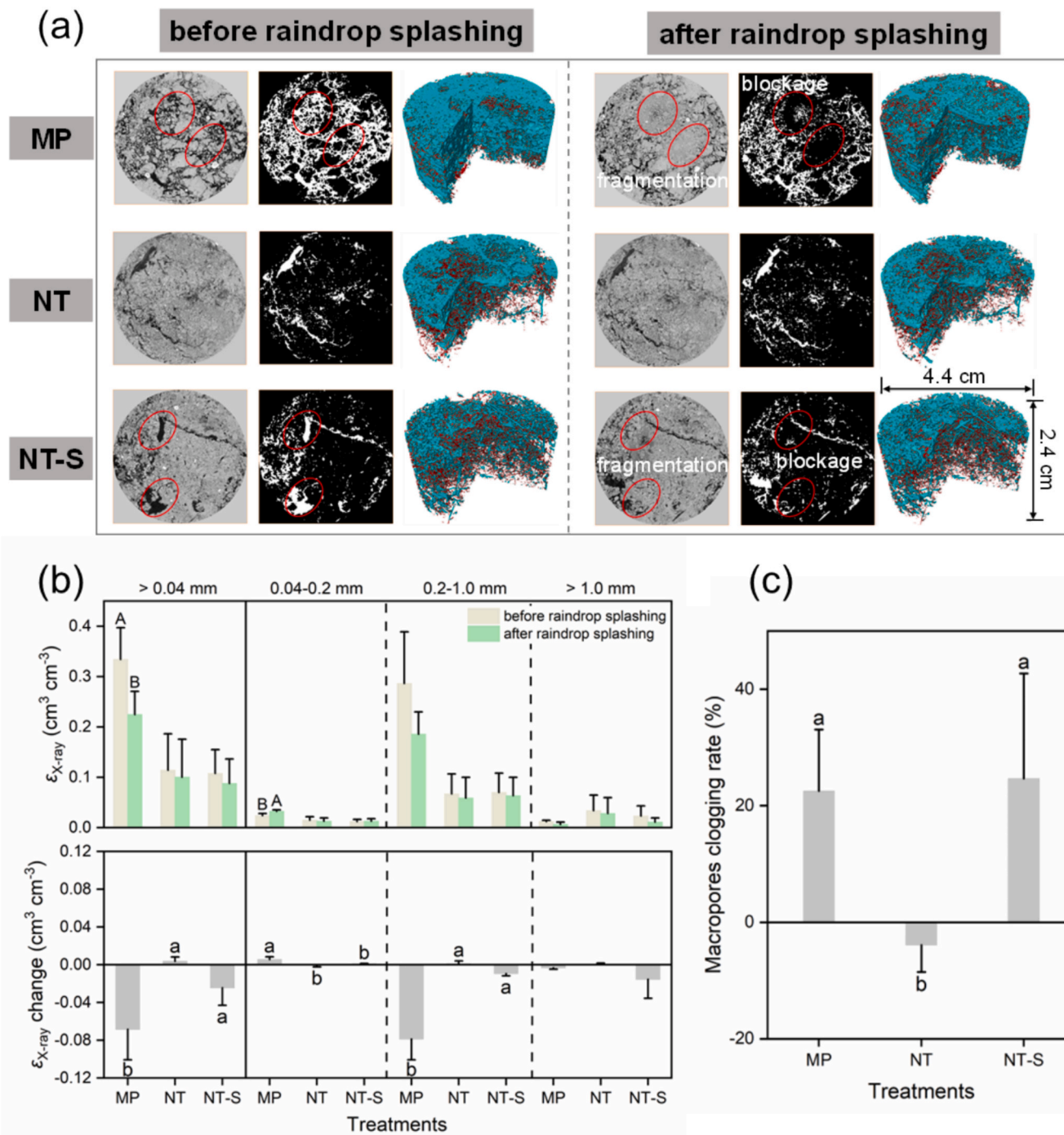


Fig. 6. The 3D visualizations of soil cores (a), macroporosity derived from X-ray CT (ϵ_{X-ray}), their changes (b), and macropores clogging rate (c) before and after raindrop splashing under moldboard plough (MP), no-tillage (NT), and no-tillage without straw mulch (NT-S) in the top part (0.1–2.4 cm). The gray color represents pores with a pore size of 40–200 μm and the cyan color represents pores with a pore size greater than 200 μm . The data is the average value ($n = 6$). Different upper-case letters indicate significant differences between before and after raindrop splashing at the same treatment ($P < 0.05$). Different lower-case letters indicate significant differences between treatments ($P < 0.05$).

3.2. Characteristics of soil pores and their changes during rainfall under contrasting tillage practices

The distribution and stability of soil pores varied significantly between NT and MP systems. While the ϵ_{total} in the 0–5 cm layer was similar between the two treatments ($P > 0.05$), MP exhibited higher ϵ_{total} in the 5–20 cm layer but lower ϵ_{total} in deeper layers (20–60 cm)

compared to NT ($P < 0.05$) (Fig. 4). X-ray-measured porosity further highlighted these contrasts: MP had 2.0 \times and 3.6 \times higher ϵ_{X-ray} than NT in the 0–5 cm and 5–10 cm layers, respectively ($P < 0.05$), but no differences were observed below 10 cm ($P > 0.05$) (Fig. 4). Macropore connectivity in MP was 63.2 % and 60.8 % higher than NT in the 5–10 and 10–20 cm layers ($P < 0.05$), whereas NT maintained greater vertical biopore abundance and higher bioporosity across most depths (Figs. 4

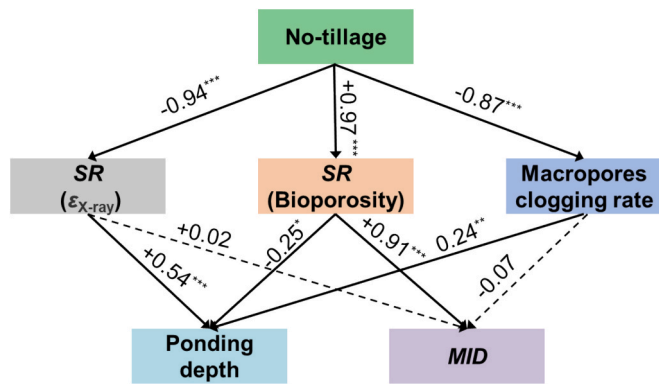


Fig. 7. Structural equation model (SEM) depicting how no-tillage affect soil infiltration capacity and how far this is mediated by soil properties. $\epsilon_{X\text{-ray}}$ is the macroporosity derived from X-ray CT. SR is the stratification ratio of $\epsilon_{X\text{-ray}}$ and bioporosity in the 0–20 and 20–60 cm layers. MID is maximum infiltration depth. Numbers next to arrows indicate standardized path coefficients. Solid lines indicate significant relationships, and dash lines indicate non-significant pathways. *, **, *** indicate significant differences at 0.05, 0.01 and 0.001 levels, respectively. The parameters of model are $\chi^2/\text{df} = 1.042$, CFI = 0.998, RMSEA = 0.062.

and 5). These structural differences were reflected in the SR value: MP had a higher SR for $\epsilon_{X\text{-ray}}$ (3.04 vs. 1.37) but a lower SR for bioporosity (0.26 vs. 0.80) compared to NT (Fig. 4), suggesting that NT fosters more stable, biologically mediated pore networks.

Raindrop splashing exerted distinct effects on pore structure, with impacts diminishing with soil depth. Below 2.4 cm, $\epsilon_{X\text{-ray}}$ remained unchanged across treatments (Fig. S5). Surface macroaggregates under MP were particularly susceptible to raindrop disruption, forming irregular depressions (Fig. 6a and S6). In the 0.1–2.4 cm layer, MP experienced a 20.4 % decline in $\epsilon_{X\text{-ray}}$ (>0.04 mm) post-rainfall ($P < 0.05$), driven by the collapse of 0.2–1.0 mm pores and a compensatory increase in 0.04–0.2 mm pores ($P < 0.05$) (Fig. 6b). In contrast, NT and NT-S showed no significant changes ($P > 0.05$), and pores >1.0 mm were unaffected in all treatments. Notably, the macropore clogging rate was lower in NT than in MP and NT-S ($P < 0.05$), though MP and NT-S did not differ significantly ($P > 0.05$) (Fig. 6c). These results demonstrate that NT mitigates raindrop-induced pore degradation, preserving infiltration capacity by maintaining larger, more stable pores.

4. Discussion

Our results demonstrated that NT significantly enhances soil infiltration capability compared to MP during heavy rainfall, as evidenced by delayed ponding onset, reduced surface water accumulation, and a greater MID (Table 1). Results from soil moisture sensors revealed that NT facilitated earlier water arrival in deeper layers (Fig. 2a), aligning with Br⁻ trace results by Hess et al. (2018) and supporting the hypothesis that NT promotes bypass flow through preferential pathways, whereas MP retains water in upper layers via matrix flow (Kulasekera et al., 2011). Dye tracer experiments further validated this result: MP exhibited higher topsoil TSA, indicative of lateral matrix flow, while NT showed deeper, vertically oriented pathways (Fig. 3).

Rainfall infiltration is primarily governed by soil pore architecture and surface conditions (Almeida et al., 2018). In this study, we evaluated pore structure by defining connected macropores as those forming continuous pathways from the top to the bottom of the soil sample (Lu et al., 2019). It is well-established that pore connectivity generally increases with $\epsilon_{X\text{-ray}}$, particularly when $\epsilon_{X\text{-ray}}$ is below a threshold of approximately $0.1 \text{ cm}^3 \text{ cm}^{-3}$ (Jarvis et al., 2017; Tian et al., 2019). While this relationship explains why the 5–20 cm layer under MP exhibited greater connected macroporosity than under NT, our findings demonstrate that infiltration capacity is controlled by the integrated

pore characteristics across the entire soil profile, not just within an individual layer.

Critically, MP management led to a highly stratified pore system with a significantly higher SR of porosity (Fig. 4). Structural equation modelling (SEM) revealed that this stratified macropore distribution under MP exacerbated surface ponding and significantly restricted the MID ($P < 0.05$) (Fig. 7). We attribute this to plow pan formation, which disrupts vertical pore continuity and promotes lateral water flow (Alletto et al., 2011; Jiang et al., 2017). In contrast, the NT system fostered a more hydrologically efficient and homogeneous pore network, characterized by greater biopore continuity and a lower macroporosity stratification ratio. The NT soil profile significantly reduced ponding and facilitated deep infiltration ($P < 0.05$) (Figs. 4 and 7). This enhanced performance is largely due to the prevalence of stable, vertical biopores formed by decayed roots and earthworm activity, which act as effective conduits for transmitting water and air through the soil profile, thereby enhancing overall permeability (Budhathoki et al., 2022; Wardak et al., 2024; Zhang et al., 2019).

Raindrop kinetic energy displaces soil particles, but its effects are tillage-dependent. MP exhibited 20.4 % greater macropore clogging in the 0.1–2.4 cm layer ($P < 0.05$), driven by aggregate breakdown into 0.04–0.2 mm pores (Fig. 6b), which compact into low-permeability crusts (Ma et al., 2022; Qin et al., 2022). NT's straw mulch mitigated this by dissipating raindrop energy, reducing splash erosion and pore clogging (Almeida et al., 2018; Wang et al., 2020). Thus, NT's dual advantages, i.e., stable surface protection and vertically connected pores, collectively enhance infiltration resilience during heavy rainfall.

In regions like Northeast China, where extreme rainfall events are increasing (Fig. S2), NT systems reduce waterlogging risks by accelerating infiltration and minimizing surface runoff. The mulch layer in NT further prevents erosion and preserves soil structure, making it a dual-purpose solution for heavy rainfall and drought (Liu et al., 2020). However, adoption requires addressing practical barriers: (1) initial yield variability during NT transition (Blanco-Canqui et al., 2017) and (2) site-specific adaptation, as NT's benefits are more pronounced in well-structured soils than in compacted or clay-rich fields (Soracco et al., 2019).

5. Conclusion

Our findings confirm that long-term no-tillage (NT) enhances infiltration capacity during extreme rainfall events through two synergistic mechanisms: (1) preservation of vertically continuous biopores that maintain profile-wide preferential flow pathways, and (2) surface protection by straw mulch that prevents pore sealing from raindrop impact. In contrast, moldboard plow (MP) systems exhibit compromised infiltration due to structural disruption of pore networks and lack of protective surface cover. These results directly support our hypothesis that NT systems are functionally superior to MP in maintaining infiltration capacity under intense precipitation.

Our findings underscore NT as a holistic soil management strategy that concurrently optimizes pore network functionality and strengthens climate resilience. By integrating structural stability with hydrological efficiency, NT systems offer a sustainable solution for rainfall-dependent agroecosystems facing increasingly erratic precipitation regimes. Future research should quantify the long-term trade-offs between tillage-induced pore evolution and crop productivity across diverse pedoclimatic contexts.

CRedit authorship contribution statement

Shijie Qin: Writing – original draft, Methodology, Investigation, Formal analysis, Data curation. **Zhuo Liu:** Writing – original draft, Formal analysis, Data curation. **Meng Tian:** Writing – review & editing, Funding acquisition. **Lin Liu:** Writing – review & editing. **Wei Hu:** Writing – review & editing. **Richard Whalley:** Writing – review & editing.

editing. **Tusheng Ren:** Writing – review & editing. **Weida Gao:** Writing – review & editing, Writing – original draft, Funding acquisition.

Declaration of competing interest

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

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Appendix A. Supplementary data

Supplementary data to this article can be found online at <https://doi.org/10.1016/j.geoderma.2025.117542>.

Data availability

Data will be made available on request.

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