Appendix A

1 Data and Methodss

1.1 Study area and time

Urban districts were also counted as counties, which can be recognized in county names by the ending "city". The Songling district and Jiagedaqiqu district were not included in Heilongjiang because their land rights belong to Inner Mongolia. Due to changes in administrative divisions, some counties were split or merged with other counties during the period of interest. As a result, some counties might be annexed, bigger or smaller in some years. The counties that were annexed were: Heilongjiang province: Acheng city (2006-2015), Hulan county (2004-2015), Shuangcheng city (2014, 2015) and Dedu county (1996-2015); Jilin province: Shuangyang county (1995-2015), Linjiang city (1990, 1991), Jiangyuan county (1990-1994, 2006-2015), Songyuan city (1990, 1991), Fuyu county (1992-1994); Liaoning province: Pulandian city (2010-2015). There were some exceptions in the statistics for facilitating the presentation of results: Mohe city (in Heilongjiang province) included Mohe county, Xinlin district and Huzhong district; Baishan city (aliased as hunjiang city in Jilin province) included Baishan city, Linjiang city and Jiangyuan county; Songyuan city (in Jilin province) included Songyuan city and Fuyu county; Panjin city (in Liaoning province) included Panjin city and Panshan county.

1.2 NUE data and covariates

The pre-processing of the initial data set before computing NUE indicators and building models were as follows: (a) filling missing data and covariates in absent years by consulting statistical yearbook/published literature, supplementing them by mean values of adjacent years (if the absent data only occurred in a few years), or by using data of neighbor counties/cities that they belong to (there was one exception: the fruit yield in Hulan county during 1990-2003, which was supplemented by Haerbin city data). For covariates, we used the poly2nb function of the spdep package (v1.1-5) to fill gaps: local downslope curvature, downslope curvature, land surface elevation, surface temperature at daytime, enhanced vegetation index in January & February and September & October, and topographic position index in Changhai county; the amount of phosphorus (by Olsen method) in Changbaichaoxianzu county, Changhai county, Ji'an city, Liaoyuan city and Suifenhe city; pH (in H2O) in Anda city, Antu county, Boli county, Changbaichaoxianzu county, Changhai county, Changling county, Dongning county, Hegang city, Helong city, Huadian city, Huanrenmanzuzizhi county, Huinan county, Jiamusi city, Ji'an city, Jiaohe county, Jingyu county, Liaoyuan city, Linjiang city, Liuhe county, Longjing city, Meihekou city, Qianan county, Qitaihe city, Shangzhi city, Suangyashan city, Suifenhe city, Tonghua city, Tonghua county, Tumen city, Wangqing county, Wushong county, Yanji city, Yanshou county and

Zhaozhou county; (b) calculating indicators which cannot be obtained directly using other available data or parameters. Detailed equations are given in Liu et al. (2020) and required parameters (N ratio of compound fertilizer, amount of N content in excretion and urine in humans and livestock, returning field rate of manure, cake ratio, amount of N content in different cake crops, returning field rate of cake fertilizer, the ratio of straw to grain, amount of N content in different straw and returning field rate of straw) used to calculate NUE indicators were presented in He et al. (2018); (c) data calibration: the final step of pre-processing and directly determining the quality of the database and accuracy of the results. For flagged suspicious values (i.e., outliers: 10 times larger or smaller than that of adjacent years), we double-checked the reference data and compared it with other available data (e.g., the economic yield must be equal to the total production divided by the planting area of the crop).

2 **Results**

2.1 Spatial and temporal variations of NUE indicators from 1990 to 2015

Spatial and temporal variations of PFP_N During the first period (1990-1995), PFP_N was very low in 32 counties (central and east Jilin province, southeast and northwest Liaoning province); low in 48 counties (within Jilin and Liaoning provinces); moderate in 37 counties (within Heilongjiang province, west Jilin province and central Liaoning province); high in 34 counties and very high in 32 counties (mostly in Heilongjiang province) (Fig. 2-B). A similar spatial pattern was found in other study periods between 1996 and 2015 (Fig. 2-C–F).

At the aggregated provincial level, the PFP_N decreased dramatically from 60 kg kg⁻¹ in 1990 to 41 kg kg⁻¹ in 2003 (with a peak of 52 kg kg⁻¹ in 1999), and then was stable until 2015 (with a small peak around 50 kg kg⁻¹ yr⁻¹ during 2011-2012) in Heilongjiang, with an average annual growth rate of -1.4% (average PFP_N of 47 kg kg⁻¹ yr⁻¹); Jilin had a moderate reduction from 32 kg kg⁻¹ in 1990 to 22 kg kg⁻¹ in 2000, followed by a fluctuation phase around 25 kg kg⁻¹ yr⁻¹, and then stayed stable at around 30 kg kg⁻¹ yr⁻¹ from 2011 to 2015, with an average annual growth rate of -0.3% (average PFP_N of 27 kg kg⁻¹ yr⁻¹); PFP_N in Liaoning went down marginally from 33 kg kg⁻¹ to 27 kg kg⁻¹ in 1994, and then rose gradually to 35 kg kg⁻¹ in 2015, with an average annual growth rate of 0.2% (average PFP_N of 34 kg kg⁻¹ yr⁻¹) (Fig. 3-A).

Spatial and temporal variations of PNB_N From 1990 to 1995, the PNB_N was very low in 41 counties (eastern Jilin, eastern and southern Liaoning); low in 62 counties (eastern Jilin and northern Liaoning); moderate in 34 counties (southern Heilongjiang, eastern and northern Jilin, northern Liaoning); high in 28 counties and very high in 18 counties (northern and

central Heilongjiang) (Fig. 5-B). A similar spatial pattern was found over the entire study period (Fig. 5-C–F).

At the aggregated provincial level, the PNB_N in Heilongjiang decreased considerably from 1.22 kg kg⁻¹ in 1990 to 0.85 kg kg⁻¹ in 2003, thereafter increasing up to 1.12 kg kg⁻¹ in 2012, and then kept steady around 1.00 kg kg⁻¹ yr⁻¹, with an average annual growth rate of -0.9% (average PNB_N of 1.02 kg kg⁻¹ yr⁻¹). Jilin had a significant reduction of PNB_N from 0.64 kg kg⁻¹ in 1990 to 0.40 kg kg⁻¹ in 2000, and then first fluctuated around 0.50 kg kg⁻¹ yr⁻¹ and later around 0.55 kg kg⁻¹ yr⁻¹ during 2010-2015, with an average annual growth rate of -0.5% (average PNB_N of 0.52 kg kg⁻¹ yr⁻¹). The PNB_N in Liaoning changed from 0.51 to 0.40 kg kg⁻¹ yr⁻¹ during 1990-2015, with an average annual growth rate of -0.40 kg kg⁻¹ yr⁻¹).

2.2 Prediction models and performance evaluation

SMLR model Pearson's correlation coefficients were calculated to quantify the pairwise relationships between the NUE indicators and explanatory variables (Appendix G). PFP_N had a significantly negative correlation with EVI in September & October, EVI in January & February, annual average daily minimum temperature, planting area index of cereal and orchard and local upslope curvature; and no significant correlation with the gross output value of agriculture (agricultural GDP) and exchangeable acidity. PFP_N had a positive correlation with nine other explanatory variables (Appendix G-A). PNB_N was significantly negatively correlated with EVI in November & December, coarse fragments (> 2 mm), planting area index of cereal and melons, agricultural GDP and Calcisols, while it was significantly positively correlated with silt content, exchangeable acidity, total phosphorus, total carbon, pH (in KCl), soil organic carbon content, available soil water capacity (with FC=pF 2.3) and planting area index of beans. Only the planting area index of tubers had no significant correlation with PNB_N (Appendix G-B).

The initial stepwise regression model for PFP_N included 27 explanatory variables, which explained 51% of the variance of PFP_N (Appendix C). Planting area index of fiber, melons and tubers, saturated water content, soil organic carbon density, ferralsols, acrisols, cambisols, cation exchange capacity (CEC) and rural population were dropped from the regression model in the VIF step to remove multicollinearity. This reduced the complexity of the model to 17 explanatory variables. The statistical significance was carried out by the confint function, and all variables were significant. For PNB_N, the initial stepwise regression model included 27 explanatory variables, which explained 71% of the variance of PNB_N (Appendix C). EVI in January & February and in May & June, planting area index of oil, the amount of watersoluble phosphorus, cambisols, gypsisols, kastanozems, histosols cumulative, available soil water capacity until wilting point and vapor pressure were dropped from the regression model in the VIF step to remove multicollinearity. This reduced the complexity of the model to 15 explanatory variables. Then statistical significance was also carried out by confint function and all variables met the requirements.

3 Discussion

3.1 Spatial and temporal variations of NUE indicators

The results indicated that the PFP_N in the counties of Heilongjiang is higher than those counties of Jilin and Liaoning. Some counties, such as Tongyu county in Jilin and Fuxinmengguzuzizhi county in Liaoning tend to obtain lower yield with higher N input, which indicates high N loss and environmental threats. Zhao et al. (2016) indicated that ecological intensification improved PFP_N from 50 kg kg⁻¹ yr⁻¹ to 66 kg kg⁻¹ yr⁻¹ in Liufangzi, Jilin province. But they only considered the influence of chemical N application on PFP_N in most cases, especially for a specific field study, where chemical fertilizer was the major N input. Therefore, their PFP_N may be overestimated compared with the actual practice. In this study, we considered the N input from both chemical N and from organic sources. This is more meaningful for analysis at the regional scale, while also more organic fertilizer was applied along with the encouragement by Chinese government for agricultural sustainability. Despite that, only 57% of counties in 2015 met the requirement of well-managed systems $(PFP_N > 60 \text{ kg kg}^{-1} \text{ yr}^{-1}$, Dobermann 2007), when considering chemical fertilizer as N input. The corresponding critical PFP_N in this study should be 35 kg kg⁻¹ yr⁻¹ when we considered both chemical and organic N inputs. In summary, there is still a huge potential to improve PFP_N in China. PFP_N in most of counties decrease during the study period, only increasing in Huma county of Heilongjiang, Beipiao city of Liaoning and Antu county of Jilin. The different PFPN trends over time could not be explained and would require more detailed information than was available.

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Appendix B Description of covariates and their sources.

Classification ¹⁾	Covariates	Abbreviation ²⁾	Description	Resolution	Unit	Source
Сгор	Enhanced vegetation index	EVI_JanFeb; EVI_MarApr; EVI_MayJun; EVI_JulAug; EVI_SepOct; EVI_NovDec	Long-term annual mean monthly (Every two month) MODIS Enhanced Vegetation Index (EVI). Derived using a stack of MOD13Q1 EVI images.	250		SoilGrids (covariates)
	Crop types	See notes	The proportion of the planting area for crop types (10 kinds of crop type) in total planting area	County	%	Municipal Yearbook (1991-2016)
Topography	Land surface elevation	DEM	DEM based on 100 m resolution from EarthEnv-DEM90. No additional	100	meter	SoilGrids (covariates)
	Terrain slope	SLPt	Terrain slope based on DEMMRG5 derived in SAGA GIS and expressed in radians.	250	radians	SoilGrids (covariates)
	Topographic position index	TPI	Topographic Position Index is the difference to the mean calculation (residual analysis) proposed by Wilson & Gallant (2000).	250	meter	SoilGrids (covariates)
	Downslope curvature	CUVd	Downslope Curvature based on DEMMRG5 derived in SAGA GIS.	250	meter	SoilGrids (covariates)
	Local upslope curvature	CUVlu	Local upslope Curvature based on DEMMRG5 derived in SAGA GIS.	250	meter	SoilGrids (covariates)
	Local downslope curvature	CUVId	Local downslope Curvature based on DEMMRG5 derived in SAGA GIS.	250	meter	SoilGrids (covariates)
Soil	Organic carbon stock	SOCs	Organic carbon stock in 0.3 m	250	tonnes	SoilGrids
	Organic carbon content	SOCc	Organic carbon content in 0.3 m	250	g kg ⁻¹	SoilGrids
	Organic carbon density	SOCd	Organic carbon density in 0.3 m	250	kg m ⁻³	SoilGrids
	Histosols cumulative probability	HP	Histosols cumulative probability in 2 m	250	%	SoilGrids
	pH in H ₂ O	PHw	pH in H ₂ O in 0.3 m	250		SoilGrids
	pH in KCl	PHk	pH in KCl in 0.3 m	250		SoilGrids

Sodic soil grade	SOD	Sodic soil grade in 2 m	250	grade	SoilGrids
Acid sub-soils grade	ACID	Acid sub-soils grade (0.15-0.6 m)	250	grade	SoilGrids
Coarse fragments	COA	Coarse fragments volumetric (>2mm) in 0.3 m	250	%	SoilGrids
volumetric					
Class content	CLAY	Taxture (alay content 0, 2 migro motor) in 0,3 m	250	0/	SoilGride
Clay content	CLAI	rexture (cray content 0-2 micro meter) in 0.5 m	230	70	Solionas
Silt content	SILT	Texture (silt content 2-50 micro meter) in 0.3 m	250	%	SoilGrids
Sand content	SAND	Texture (sand content 50-2000 micro meter) in 0.3 m	250	%	SoilGrids
Bulk density	BD	Bulk density in 0.3 m	250	kg m ⁻³	SoilGrids
Absolute depth to bedreak	DED	Absolute danth to hadroak	250		Shangeyon at al. 2014
Absolute deput to bedrock	DED	Absolute depth to bedrock	250	ciii	Shangguan et al, 2014
Soil types	SOIL	Soil types (17 kinds of soil type)	250	% or grade	SoilGrids
				Sidde	
The amount of phosphorus	POL		1000	ppm of	Shangguan et al, 2014
by Olsen method				weight	
The amount of water	РНО		1000	ppm of	Shangguan et al, 2014
soluble phosphorus				weight	
Total carbon	тс		1000	% of	Shangguan et al. 2014
Total carbon			1000	weight	Shangguan et al, 2014
Total nitrogen	TN		1000	% of	Shangguan et al, 2014
Total phosphorus	ТР		1000	weight % of	Shangguan et al. 2014
				weight	
Total potassium	ТК		1000	% of weight	Shangguan et al, 2014
Total sulfur	TS		1000	% of	Shangguan et al, 2014
				weight	

Exchangeable aluminum	EXAL		1000	cmol kg ⁻¹	Shangguan et al, 2014
Exchangeable calcium	EXCA		1000	cmol	Shangguan et al, 2014
Exchangeable magnesium	EXMG		1000	cmol	Shangguan et al, 2014
Exchangeable potassium	EXK		1000	cmol	Shangguan et al, 2014
Exchangeable sodium	EXNA		1000	cmol	Shangguan et al, 2014
Exchangeable acidity	EXH		1000	cmol	Shangguan et al, 2014
Cation exchange capacity	CEC	Cation exchange capacity in 0.3 m	250	cmol kg ⁻¹	SoilGrids
Electrical conductivity	EC		1000	ds m ⁻¹ or mmho	Shangguan et al, 2014
Available soil water capacity with FC=pF 2.0	WATal	Available soil water capacity with FC=pF 2.0 in 0.3 m	250	cm ⁻¹ %	SoilGrids
Available soil water capacity with FC=pF 2.3	WATa2	Available soil water capacity with FC=pF 2.3 in 0.3 m	250	%	SoilGrids
Available soil water capacity with FC=pF 2.5	WATa3	Available soil water capacity with FC=pF 2.5 in 0.3 m	250	%	SoilGrids
Available soil water capacity (volumetric fraction) until wilting point	WATa	Available soil water capacity (volumetric fraction) until wilting point in 0.3 m	250	%	SoilGrids

Saturated water content (volumetric fraction)	WATs	Saturated water content (volumetric fraction) in 0.3 m	250	%	SoilGrids
Annual average surface temperature (daytime)	TMPd	Long-term averaged mean annual surface temperature (daytime) MODIS. Derived using a stack of MOD11A2 LST images.	1000	Kelvin	SoilGrids (covariates)
Annual average surface temperature (nighttime)	TMPn	Long-term averaged mean annual surface temperature (nighttime) MODIS. Derived using a stack of MOD11A2 LST images.	1000	Kelvin	SoilGrids (covariates)
Daily mean temperature	TMP		10000	°C	Climatic Research Unit
Annual average daily maximum temperature	TMX		10000	°C	Climatic Research Unit
Annual average daily	TMN		10000	°C	Climatic Research Unit
Diurnal temperature range	DTR		10000	°C	Climatic Research Unit
Cloud cover	CLD		10000	%	Climatic Research Unit
Potential evapo-	PET		10000	Millim	Climatic Research Unit
transpiration Vapour pressure	VAP		10000	etres Hecta- Pascals	Climatic Research Unit
Wet day frequency	WET	Wet day frequency (rain days per year)	10000	Days	Climatic Research Unit
Precipitation	PRE		10000	Millim	Climatic Research Unit
Frost day frequency	FRS		10000	Days	Climatic Research Unit
Gross output value of agriculture	GDP		County	RMB ha ⁻¹	Municipal Yearbook (1991-2016)
Rural population	POP	Rural population per planting area	County	person ha ⁻¹	Municipal Yearbook (1991-2016)
Available irrigation area index	IRRI	Available irrigation area per planting area		ha ha ⁻¹	Municipal Yearbook (1991-2016)
	Saturated water content (volumetric fraction) Annual average surface temperature (daytime) Annual average surface temperature (nighttime) Daily mean temperature Annual average daily maximum temperature Diurnal average daily minimum temperature Diurnal temperature range Cloud cover Potential evapo- transpiration Vapour pressure Wet day frequency Precipitation Frost day frequency Gross output value of agriculture Rural population Available irrigation area index	Saturated water content (volumetric fraction)WATsAnnual average surface temperature (daytime)TMPdAnnual average surface temperature (nighttime)TMPnDaily mean temperatureTMPAnnual average daily maximum temperatureTMNAnnual average daily minimum temperatureTMNOurnal average daily minimum temperatureTMNOurnal temperature Diurnal temperature rangeDTRCloud coverCLDPotential evapo- transpiration Vapour pressurePETVapour pressureVAPWet day frequencyWETPrecipitationPREFrost day frequencyGDPAvailable irrigation area indexIRRI	Saturated water content (volumetric fraction)WATsSaturated water content (volumetric fraction) in 0.3 mAnnual average surface temperature (daytime)TMPdLong-term averaged mean annual surface temperature (daytime) MODIS. Derived using a stack of MOD11A2 LST images.Annual average surface temperature (nighttime)TMPnLong-term averaged mean annual surface temperature (nighttime) MODIS. Derived using a stack of MOD11A2 LST images.Daily mean temperature maximum temperatureTMPLong-term averaged mean annual surface temperature (nighttime) MODIS. Derived using a stack of MOD11A2 LST images.Annual average daily maximum temperature Diumal temperature Durmal temperatureTMNCloud coverCLDPotential evapo- transpiration Vapour pressurePET TET TET TETPrecipitation agriculturePREFrost day frequency Rural populationGDP Rural population per planting areaAvailable irrigation area agricultureIRRIAvailable irrigation area per planting area	Saturated water content (volumetric fraction) WATs Saturated water content (volumetric fraction) in 0.3 m 250 Annual average surface temperature (daytime) TMPd Long-term averaged mean annual surface temperature (daytime) MODIS. Derived using a stack of MOD11A2 LST images. 1000 Annual average surface temperature (nighttime) TMPn Long-term averaged mean annual surface temperature (nighttime) MODIS. Derived using a stack of MOD11A2 LST images. 1000 Daily mean temperature famperature (nighttime) TMP Long-term averaged mean annual surface temperature (nighttime) MODIS. 10000 Annual average daily maximum temperature Diurnal temperature tumperature reperature plurnal temperature transpiration Vapour pressure TMN 10000 Cloud cover CLD 10000 10000 Vet day frequency WET Wet day frequency (rain days per year) 10000 Precipitation PRE 10000 10000 Frost day frequency FRS 10000 Gross output value of agriculture GDP County Rural population POP Rural population area per planting area County	Saturated water content (volumetric fraction) WATs Saturated water content (volumetric fraction) in 0.3 m 250 % Annual average surface temperature (daytime) TMPd Long-term averaged mean annual surface temperature (daytime) MODIS. Derived using a stack of MOD11A2 LST images. 1000 Kelvin Annual average surface temperature (nighttime) TMPn Long-term averaged mean annual surface temperature (nighttime) MODIS. 1000 °C Daily mean temperature (nighttime) TMPn Long-term averaged mean annual surface temperature (nighttime) MODIS. 10000 °C Annual average daily maximum temperature Diurnal temperature annual average daily minimum temperature Durnal temperature ange DTR TMN 10000 °C Cloud cover CLD 10000 % % Potential evapo- transpiration PET 10000 % Vapour pressure VAP 10000 Hetta- Pascali Wet day frequency WET Wet day frequency (rain days per year) 10000 Days Precipitation PRE 10000 Days Days Frost day frequency FRS County RMB ha ⁻¹ Rural population area index IRRI Available irrigation area per planting area County Parson ha ⁻¹

¹⁾ AMP: agricultural management practice.

²⁾ 17 soil types: Acrisols, ACR; Albeluvisols, ALB; Alisols, ALI; Andosols, AND; Arenosols, ARE; Calcisols, CAL; Cambisols, CAM; Chernozems, CHE; Cryosols, CRY; Durisols, DUR; Ferralsols, FER; Fluvisols, FLU; Gleysols, GLE; Gypsisols, GYP; Histosols, HIS; Kastanozems, KAS; Leptosols, LEP. 10 crop types: Cereal, CER; Beans, BEAN; Tubers, TUB; Oil, OIL; Sugar, SUG; Fiber, FIB; Tobacco, TOB; Vegetables, VEG; Orchards, ORC; Melons, MEL. Climatic Research Unit: (<u>http://www.cru.uea.ac.uk/data</u>); Shangguan, W., Dai, Y., Duan, Q., Liu, B., & Yuan, H. (2014). A global soil data set for earth system modeling. Journal of Advances in Modeling Earth Systems, 6(1), 249-263; SoilGrids. (<u>https://soilgrids.org/</u>).



Appendix C The stepwise linear regression model building procedure. The R functions that were used for building the model are shown in the purple shaded arrows. MEC refers to the model efficiency coefficient of the regression models. VIF means variance inflation factor. The numbers under "covariates" represent the number of covariates in the model, and the percentage numbers under the MEC indicate how much MEC is explained by the model.



Appendix D Spatial distribution of average yield (A), nitrogen input (B) and crop removal (C) from 1990 to 2015.



Appendix E Cumulative MSE (A, D), RMSE (B, E) and MEC (C, F) (scatter plot: x axis- trees) in Random Forest models for the PFP_N and PNB_N , respectively.



Appendix F Relative importance of explanatory variables for the PFP_N : A) LMG method in in calc.relimp function; B) permutation method and C) impurity methods for RF model with ranger function; and for PNB_N : D) LMG method in calc.relimp function; E) permutation method and F) impurity methods for RF model with ranger function. Note: See Abbreviations (Appendix B) for explanations of the variable code names.



Appendix G Pairwise comparison of all explanatory variables used in the PFP_N (A) and PNB_N (B) regression models. The upper-right panel contains predicted pair-wise Pearson's correlations. The diagonal panel shows histograms and the lower-left panel shows scatterplots with a LOESS smoother added to aid visual interpretation. Top row and left columns show bivariate relations between yield and explanatory variables. Note: See Abbreviations (Appendix B) for explanation of the variable code names. *, *P-value*<0.05; **, *P-value*<0.01; ***, *P-value*<0.001.