

Supplement of

Dynamics of China's tree canopy cover: a China annual tree canopy cover dataset at 30m resolution from 1970s to 2024

Jinge Yu¹, Zhen Yu^{1,2,3,4 *}, Wangya Han^{1,4}, Fangmin Zhang^{1,4}, Shirong Liu^{2,3 *}

¹State Key Laboratory of Climate System Prediction and Risk Management, Nanjing University of Information Science and Technology, Nanjing, 210044, China

²Institute of Forestry and Grassland Carbon Sequestration, Chinese Academy of Forestry, Beijing, 100091, China

³Key Laboratory of Forest Ecology and Environment, China's National Forestry and Grassland Administration, Ecology and Nature Conservation Institute, Chinese Academy of Forestry, Beijing, 100093, China

⁴Key Laboratory of Ecosystem Carbon Source and Sink, China Meteorological Administration (ECSS-CMA), School of Ecology and Applied Meteorology, Nanjing University of Information Science and Technology, Nanjing, 210044, China

**Correspondence to:* Zhen Yu (zyu@nuist.edu.cn) & Shirong Liu (liusr@caf.ac.cn)

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Supplementary

Table (S1–S5)

Table S1. Overview of datasets used for model input variables.

Dataset	Regions	Temporal scope	Spatial resolution	Sources
Landsat	China	1985-2024	30 m	(Cai et al., 2025)
ALCC	China	1985-2022	30 m	(Li et al., 2026)
NASA DEM	Global	2000	30 m	https://www.earthdata.nasa.gov/data/catalog/lpcloud-nasadem-hgt-001
TerraClimate	Global	1985-2024	4638.30 m	(Abatzoglou et al., 2017)

Table S2. Overview of tree canopy cover relate-datasets utilized for comparison.

Dataset	Regions	Temporal scope	Spatial resolution	Sources
MODIS VCF	Global	2000-2020 (Annual)	250 m	https://lpdaac.usgs.gov/products/mod44bv061/
GFCC	Global	2000, 2010, 2015	30 m	(Sexton et al., 2013)
DWTP	Global	2015-2026	10 m	(Brown et al., 2022)
Hansen's GFC	Global	2000	30 m	(Hansen et al., 2013)
CATCD	China	1985-2023	30 m	(Cai et al., 2024)

Table S3. Overview of forest definition datasets used for area comparison.

Dataset	Regions	Temporal scope	Spatial resolution	Sources
Land cover/Land use datasets				
ALCC	China	1985-2022	30 m	(Li et al., 2026)
CLCD	China	1990-2024 (Annual)	30 m	(Yang and Huang, 2021)
GlobeLand30	Global	2000, 2010, 2020	30 m	(Chen et al., 2015)
GLC_FCS30D	Global	1985, 1990, 1995, 2000-2022	30 m	(Zhang et al., 2024)
Datasets Related to tree canopy cover				
MODIS VCF	Global	2000-2020 (Annual)	250 m	https://lpdaac.usgs.gov/products/mod44bv061/
GFCC	Global	2000, 2010, 2015, 2020	30 m	(Sexton et al., 2013)
Hansen's GFC	Global	2000, 2010	30 m	(Hansen et al., 2013)
CATCD	China	1985-2023	30 m	(Cai et al., 2024)

Table. S4. Comparison of candidate models for stand age–stem density relationships across species.

Species	Model	Equation	R ²	Adj_R ²	AIC	p_value
<i>Picea</i> spp.	Exp.	$y = 3248.8 \cdot e^{-0.011x} + 94.7$	0.49	0.49	9398.27	<0.001
	Pow.	$y = 49825.2 \cdot x^{-0.829}$	0.46	0.46	9435.14	<0.001
	Log.	$y = 6260.9 - 1105.055 \cdot \ln(x)$	0.49	0.49	9399.47	<0.001
<i>Larix</i> spp.	Exp.	$y = 6203.4 \cdot e^{-0.024x} + 557.9$	0.59	0.59	11053.82	<0.001
	Pow.	$y = 100000 \cdot x^{-0.964}$	0.60	0.59	11037.45	<0.001
	Log.	$y = 8214.1 - 1508.9 \cdot \ln(x)$	0.56	0.56	11099.35	<0.001
<i>Pinus sylvestris</i> var. <i>mongolica</i>	Exp.	$y = 13854.4 \cdot e^{-0.065x} + 387.8$	0.42	0.40	1033.94	<0.001
	Pow.	$y = 100000 \cdot x^{-1.158}$	0.38	0.35	1039.45	<0.001
	Log.	$y = 6254.7 - 1309.52 \cdot \ln(x)$	0.34	0.32	1042.48	<0.001
<i>Pinus tabulaeformis</i>	Exp.	$y = 12603.4 \cdot e^{-0.061x} + 606.2$	0.51	0.50	5367.85	<0.001
	Pow.	$y = 100000 \cdot x^{-1.114}$	0.48	0.47	5394.11	<0.001
	Log.	$y = 7250.6 - 1512.494 \cdot \ln(x)$	0.44	0.43	5422.87	<0.001
<i>Pinus armandii</i>	Exp.	$y = 12259 \cdot e^{-0.062x} + 273.9$	0.42	0.39	767.73	<0.001
	Pow.	$y = 100000 \cdot x^{-1.164}$	0.37	0.34	772.85	<0.001
	Log.	$y = 9011 - 2069.26 \cdot \ln(x)$	0.38	0.36	769.59	<0.001
<i>Pinus densata</i>	Exp.	$y = 15000 \cdot e^{-0.059x} + 488.6$	0.53	0.51	1372.16	<0.001
	Pow.	$y = 100000 \cdot x^{-1.1}$	0.46	0.44	1386.82	<0.001
	Log.	$y = 7823.9 - 1634.779 \cdot \ln(x)$	0.43	0.42	1390.18	<0.001
<i>Cupressus funebris</i>	Exp.	$y = 3662.4 \cdot e^{-0.013x} + 338.8$	0.52	0.52	4640.21	<0.001
	Pow.	$y = 44397.5 \cdot x^{-0.773}$	0.51	0.50	4651.27	<0.001
	Log.	$y = 6488.5 - 1096.409 \cdot \ln(x)$	0.51	0.51	4647.26	<0.001
<i>Quercus</i> spp.	Exp.	$y = 8128.2 \cdot e^{-0.032x} + 594.8$	0.58	0.58	17031.05	<0.001
	Pow.	$y = 100000 \cdot x^{-0.983}$	0.57	0.57	17079.91	<0.001
	Log.	$y = 9678 - 1901.524 \cdot \ln(x)$	0.54	0.54	17148.55	<0.001
<i>Betula platyphylla</i>	Exp.	$y = 15000 \cdot e^{-0.057x} + 727.9$	0.44	0.43	8920.93	<0.001
	Pow.	$y = 100000 \cdot x^{-1.054}$	0.39	0.39	8976.24	<0.001
	Log.	$y = 9115.2 - 1893.318 \cdot \ln(x)$	0.36	0.36	9007.11	<0.001

Note: Three candidate models were used to fit the self-thinning process of tree species under similar site conditions. Exp., exponential model; Pow., power model; Log., logarithmic model.

Table S5. Independent test-set validation of plot-scale empirical equations describing stand age–stem density relationships.

Species	Equation	R ²	r	Bias
<i>Picea</i> spp.	$y = 3248.8 \cdot e^{-0.011x} + 94.7$	0.65	0.81	16.94
<i>Larix</i> spp.	$y = 100000 \cdot x^{-0.964}$	0.58	0.80	82.56
<i>Pinus sylvestris</i> var. <i>mongolica</i>	$y = 13854.4 \cdot e^{-0.065x} + 387.8$	0.61	0.80	116.73
<i>Pinus tabuliformis</i>	$y = 12603.4 \cdot e^{-0.061x} + 606.2$	0.66	0.82	23.52
<i>Pinus armandii</i>	$y = 12259 \cdot e^{-0.062x} + 273.9$	0.68	0.83	40.71
<i>Pinus densata</i>	$y = 15000 \cdot e^{-0.059x} + 488.6$	0.69	0.84	7.78
<i>Cupressus funebris</i>	$y = 3662.4 \cdot e^{-0.013x} + 338.8$	0.57	0.80	74.66
<i>Quercus</i> spp.	$y = 8128.2 \cdot e^{-0.032x} + 594.8$	0.65	0.81	30.91
<i>Betula platyphylla</i>	$y = 15000 \cdot e^{-0.057x} + 727.9$	0.59	0.77	-4.68

Figure (S1–S8)

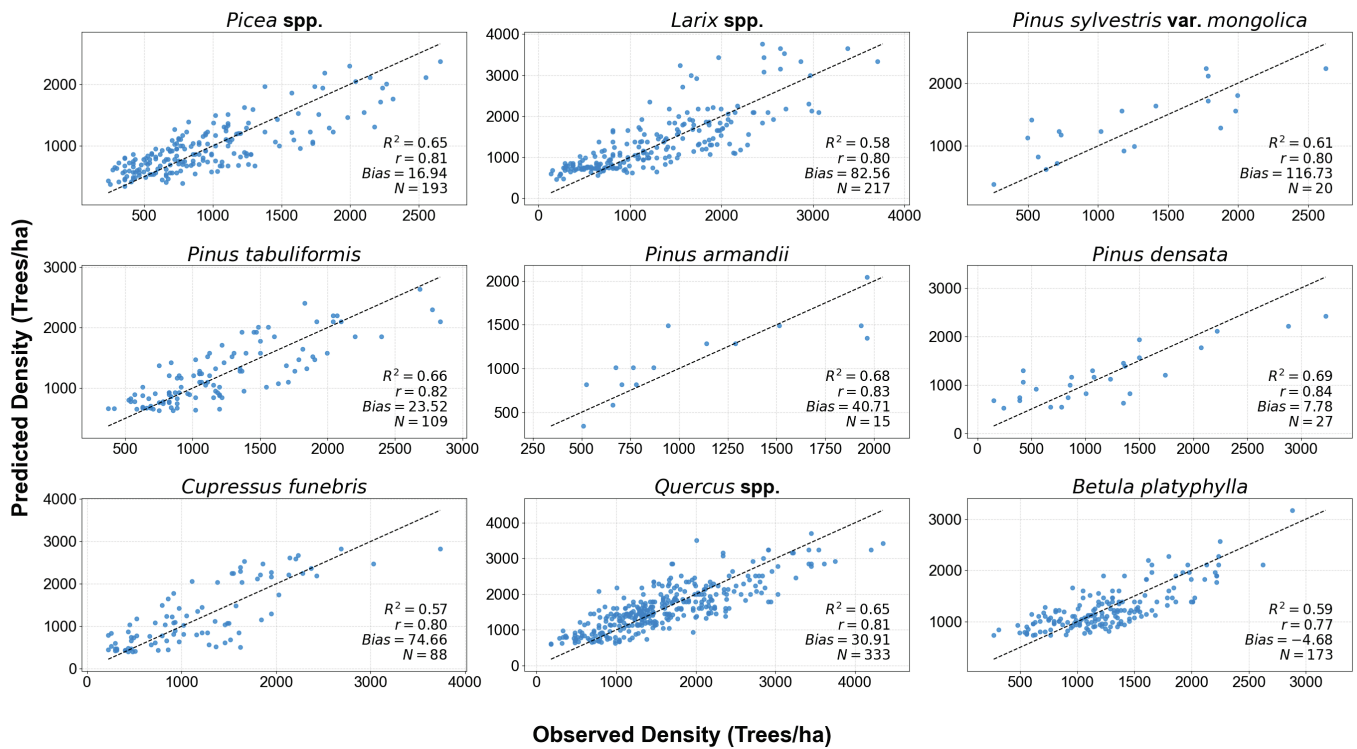


Figure S1. Comparison between observed and predicted stem density for each species.

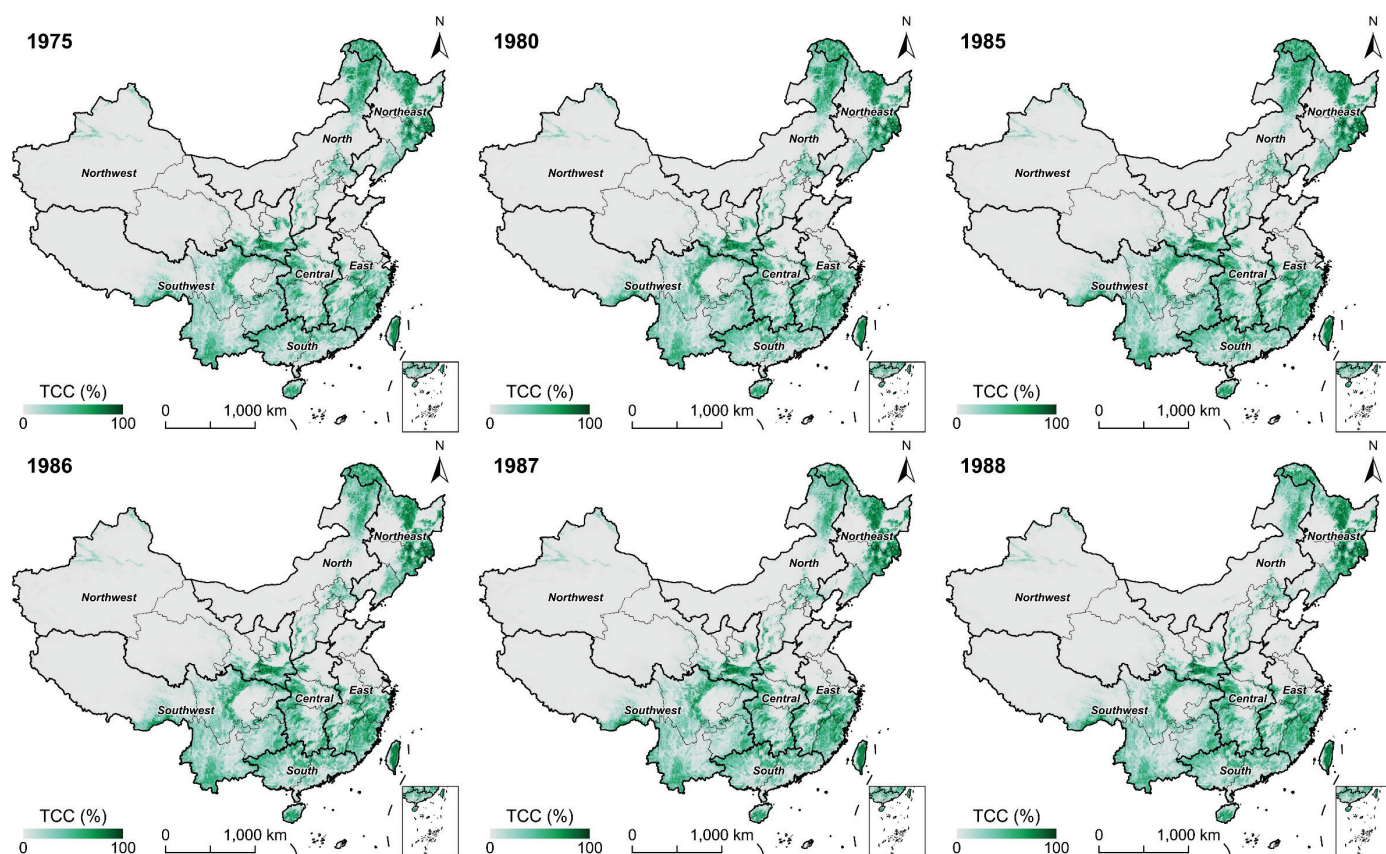


Figure S2. Spatial distribution of ChinaTCC30 in China for 1975, 1980, 1985-1988.

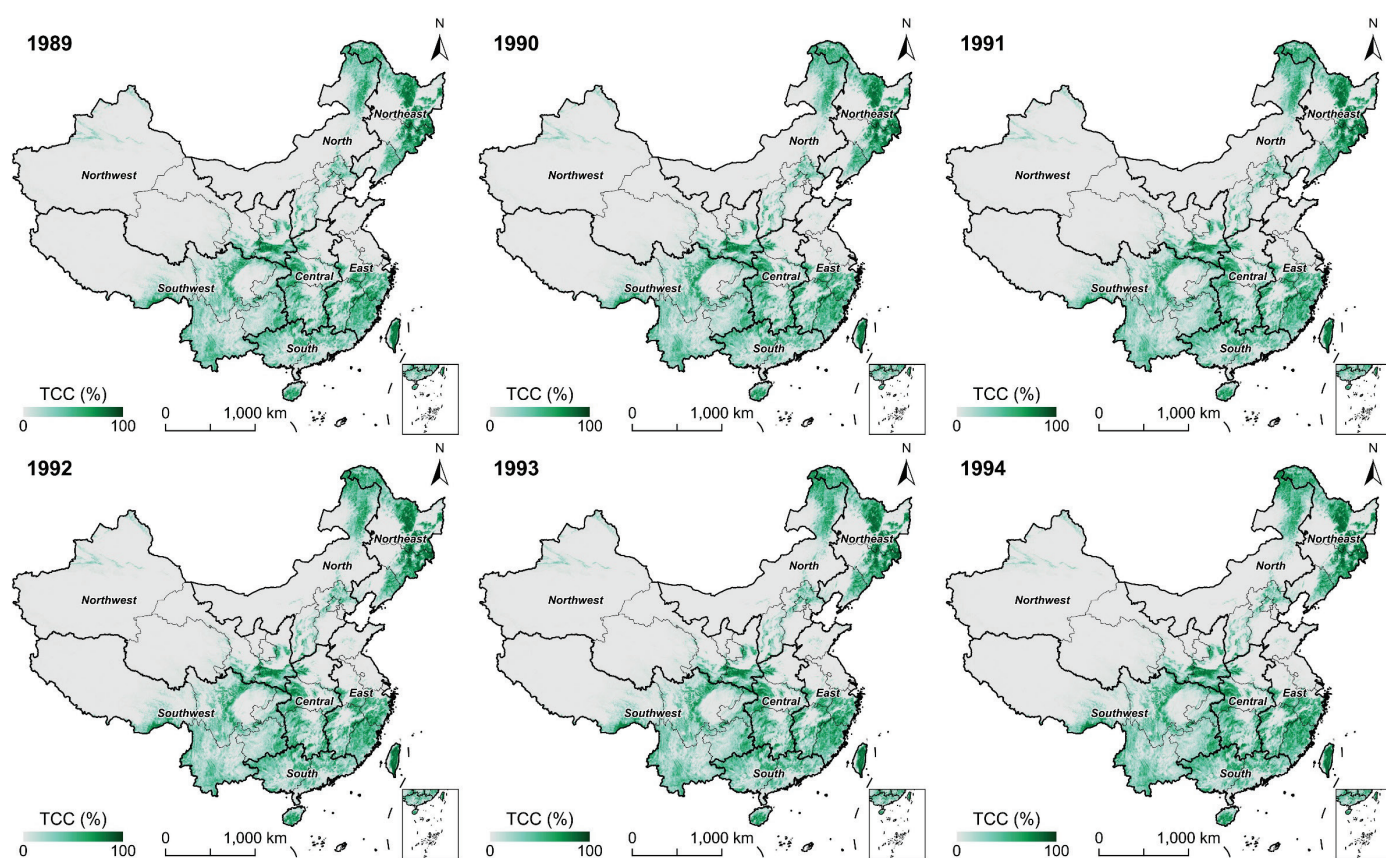


Figure S3. Spatial distribution of ChinaTCC30 in China for 1989-1994.

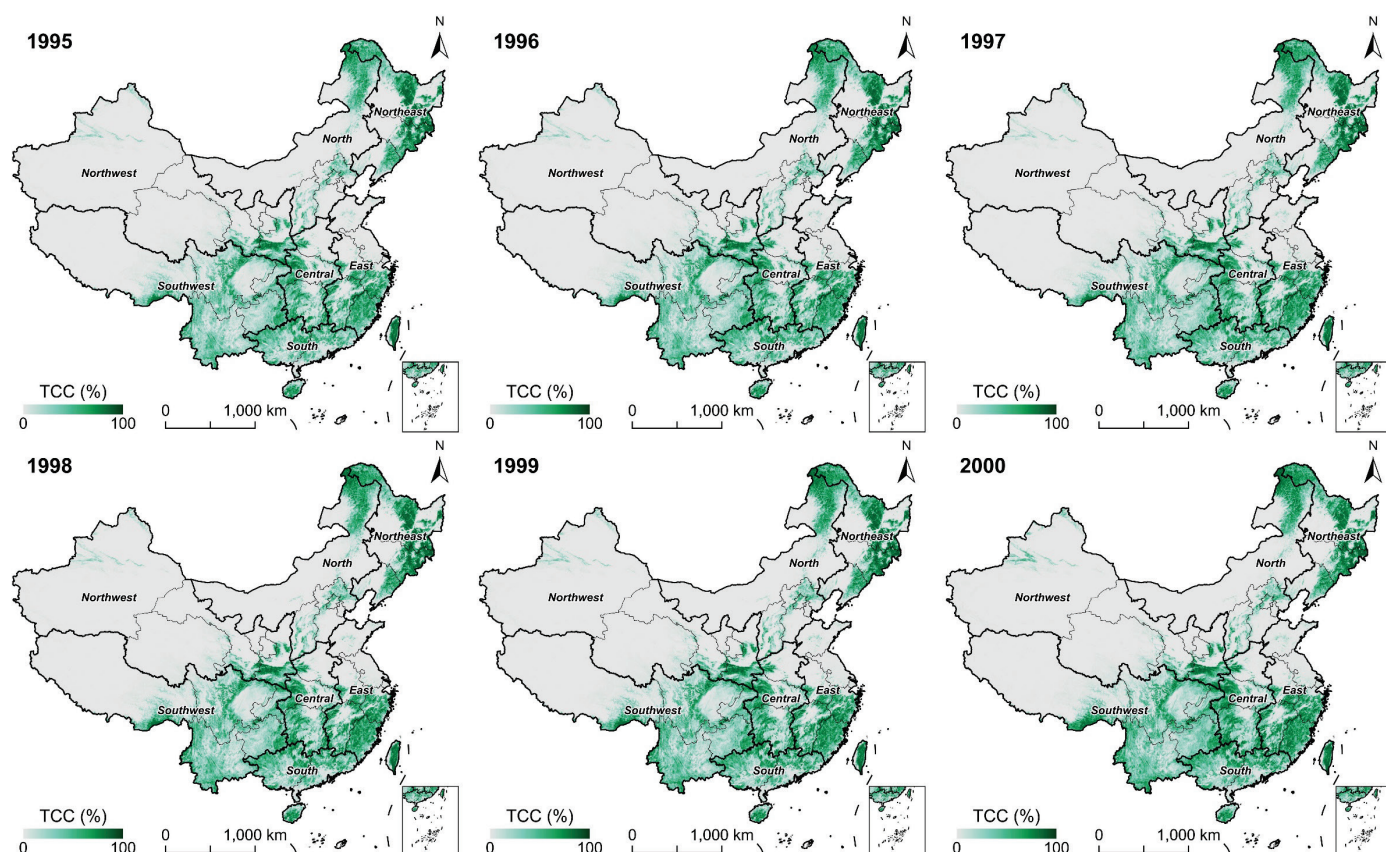


Figure S4. Spatial distribution of ChinaTCC30 in China for 1995-2000.

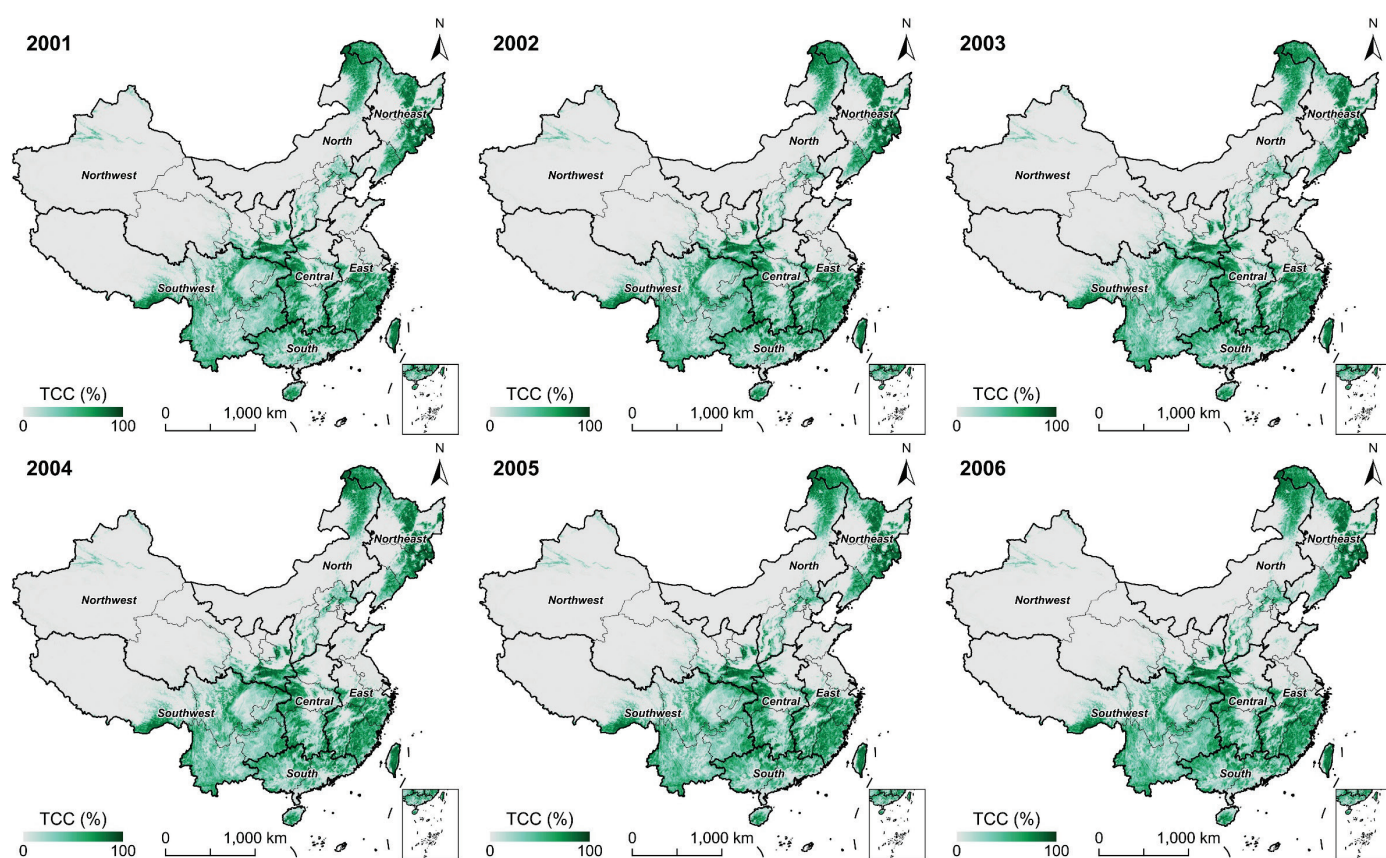


Figure S5. Spatial distribution of ChinaTCC30 in China for 2001-2006.

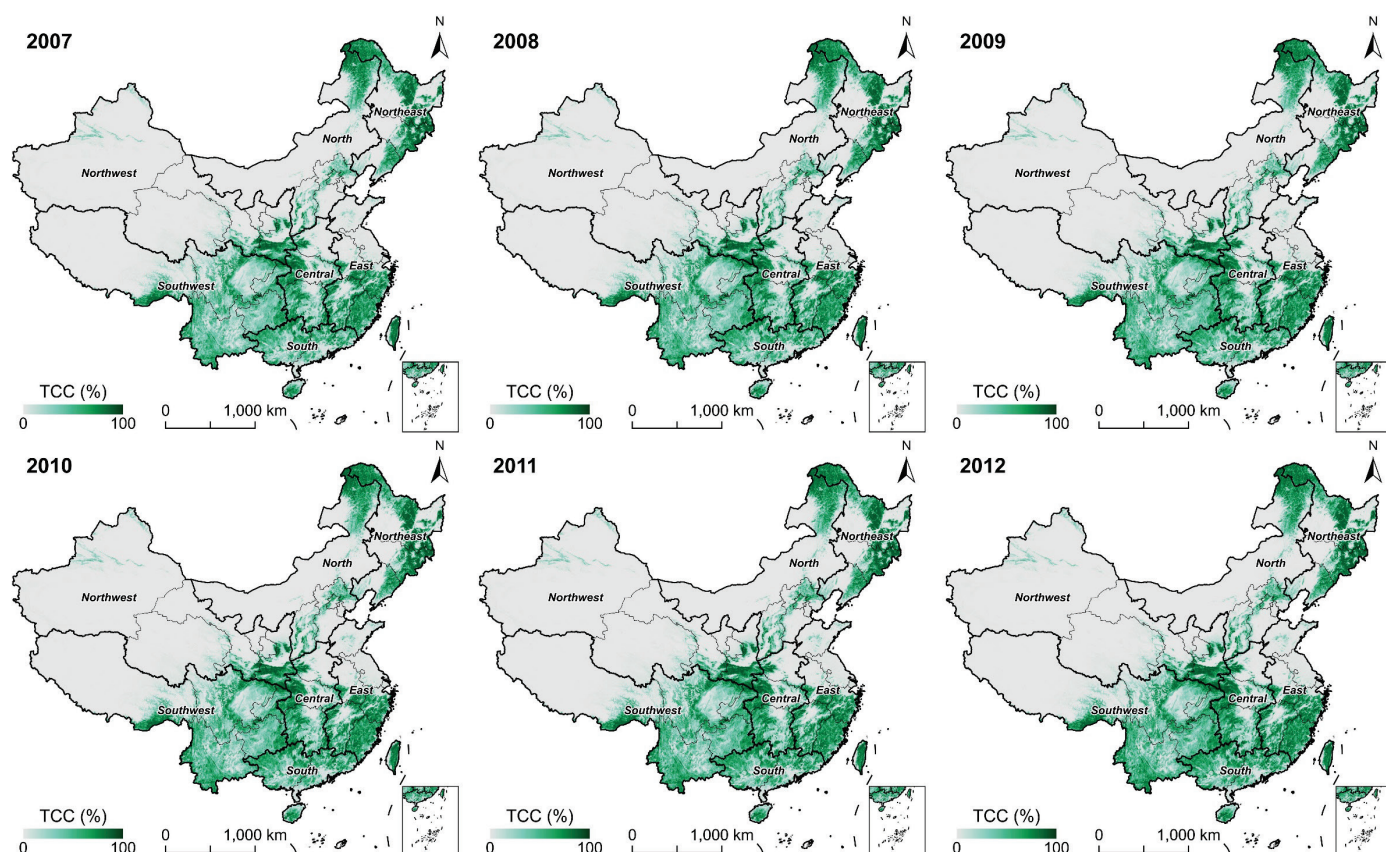


Figure S6. Spatial distribution of ChinaTCC30 in China for 2007-2012.

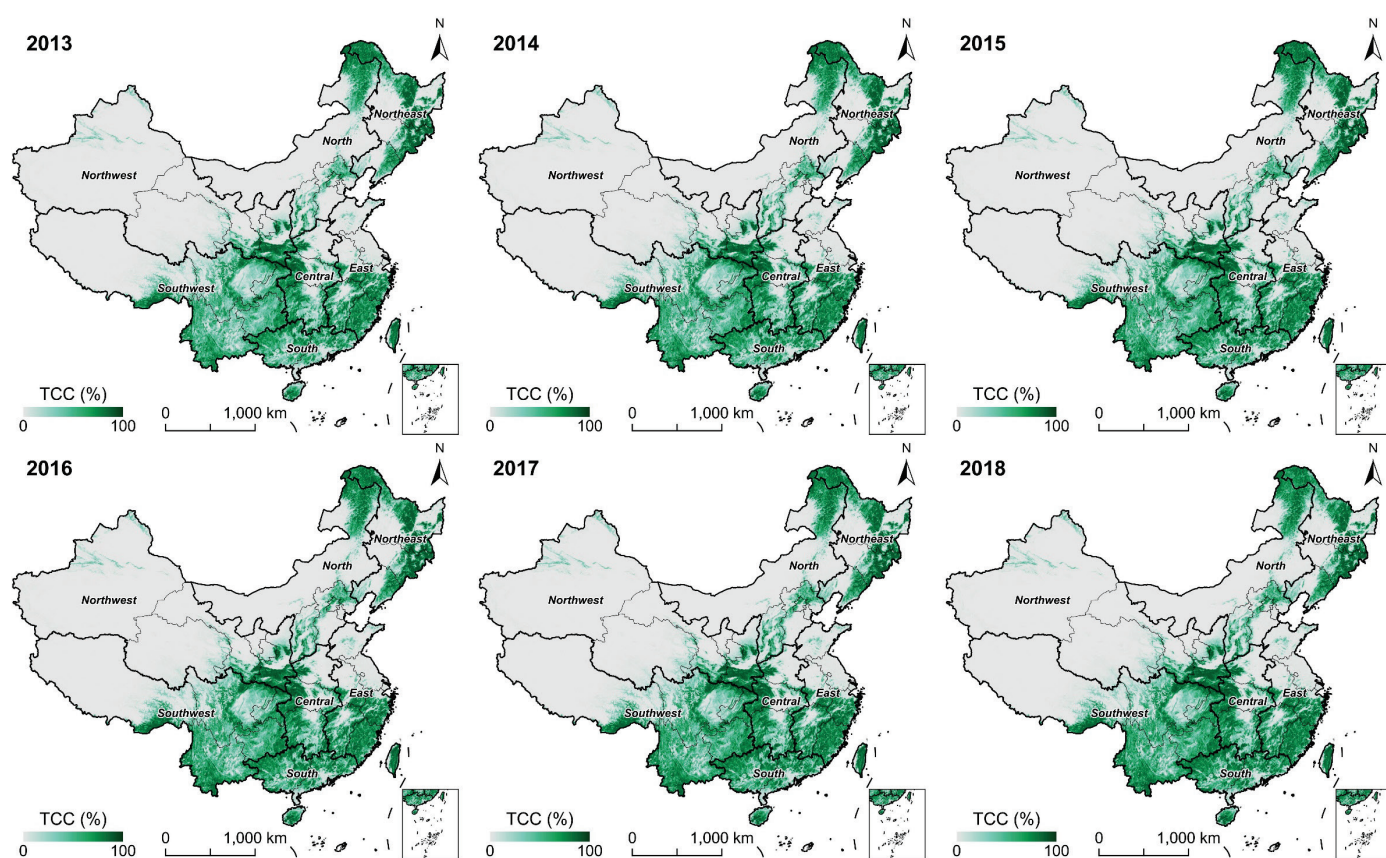


Figure S7. Spatial distribution of ChinaTCC30 in China for 2013-2018.

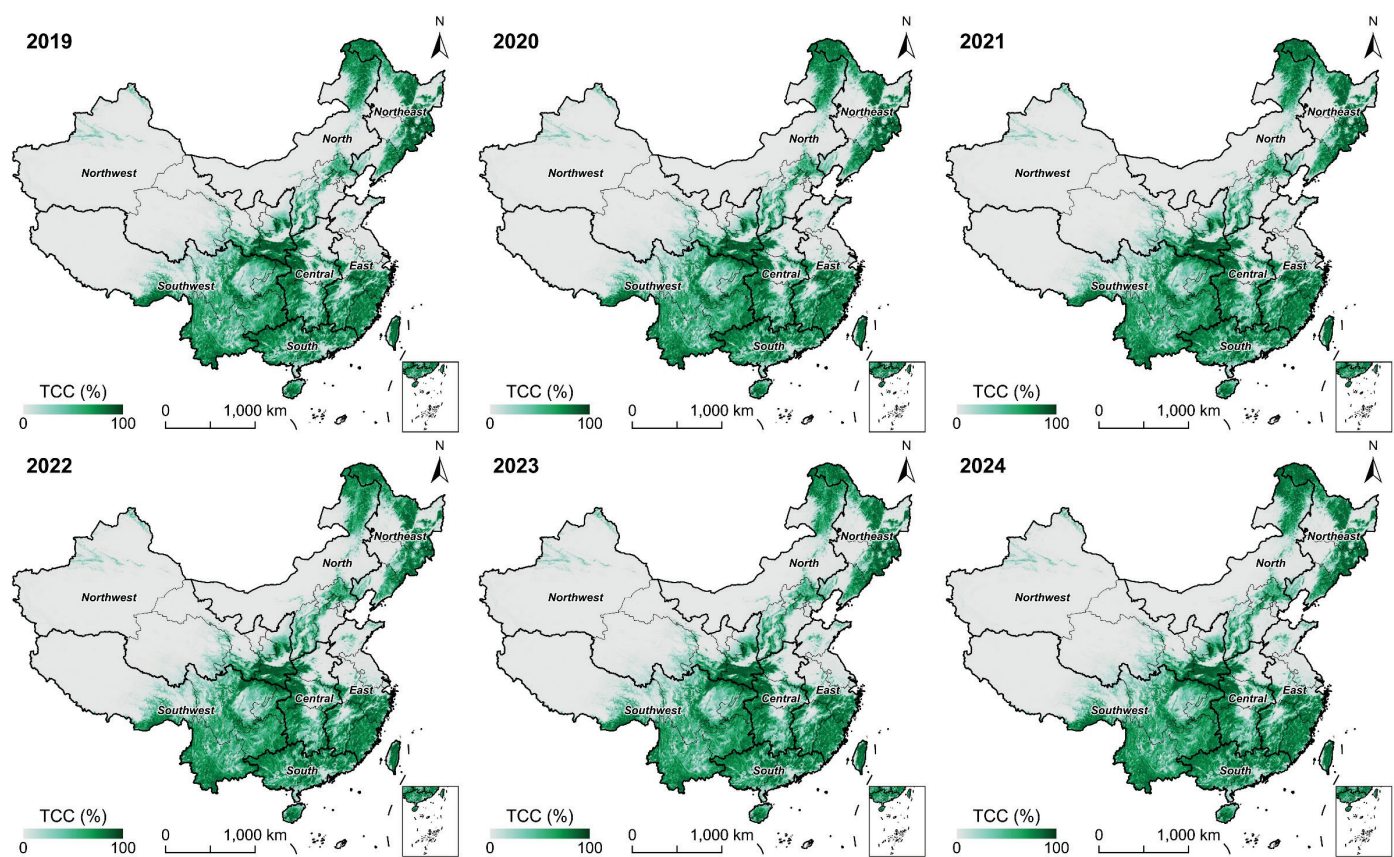


Figure S8. Spatial distribution of ChinaTCC30 in China for 2019-2024.

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