Supplementary Information for

A 30 m annual cropland dataset of China from 1986 to 2021

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Figure S1. Divisions of the study area. Annual cropland classification is performed within each $0.8^{\circ} \times 0.8^{\circ}$ subregion. Test regions with a size of 100 km×100 km are used to find the best LandTrendr arguments for each agricultural zone.

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Figure S2. An example of the sample interpretation process using the developed Cropland Inspector Tool on © Google Earth Engine (<u>https://code.earthengine.google.com/616d493de325aaa5e3316a58ed8e7531</u>). The location was covered by croplands before 2009 but converted to built-up areas since then, of which changes were clearly shown in Landsat images and NDVI time series.



Figure S3. Mean F1 score of annual cropland classification results under different training sample sizes.



Figure S4. A comparison of a true color Landsat imagery displayed (left) and a corresponding NDVI composite imagery (with 10th, 50th, and 90th percent quantile of one-year values as the RGB channels) on the right. The Landsat imagery is provided by USGS with free access.



Figure S5. Distribution of global land cover validation sample set (GLCVSS) in China.



Figure S6. Distribution of GeoWiki cropland samples in China.



Figure S7. Provincial cropland abandonment in China between 1990-2015 of (a) area and (b) rate.

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Table S1. A summary of the nine agricultural zones in China: their geographical and climatic conditions, growing seasons, cropping patterns, and major crops

Category	Feature	Description	Dimension	Source
Spectrum	10 th , 25 th , 50 th , 75 th ,	Spectral bands of Landsat data	6*5*2	Landsat
	and 90 th percent			
	quantiles of Blue,			
	Green, Red, NIR,			
	SWIR1, and SWIR2			
	bands of both			
	growing and non-			
	growing seasons			
Spectral	10 th , 25 th , 50 th , 75 th ,	Normalized indices derived from	5*5*2	Landsat
indices	and 90 th percent	spectral bands, which are		
	quantiles of NDVI,	calculated as:		
	NDSI, NDBI, NBR,	NDVI = (NIR-Red)/(NIR+Red)		
	and EVI indices of	NDSI = (Green-		
	both growing and	SWIR1)/(Green+SWIR1)		
	non-growing seasons	NDBI = (SWIR1-		
		NIR)/(SWIR1+NIR)		
		NBR = (NIR-		
		SWIR1)/(NIR+SWIR2)		
		EVI = 2.5*((NIR-		
		Red)/(NIR+6*Red-7.5*Blue+1))		
Tasseled cap	10 th , 25 th , 50 th , 75 th ,	Tasseled cap transformation	3*5*2	
transformation	and 90 th percent	indices of spectral bands.		
indices	quantiles of	Coefficients are derived from		
	brightness,	Crist (1985).		
	greenness, and			
	wetness indices of			
	both growing and			
	non-growing seasons			
Topography	Elevation	/	1	SRTM

Table S2. Input features of multi-temporal metrics each year for the random forest classifier for estimating annual cropland probabilities.

Noted NIR and SWIR are short for the near-infrared and short-wave bands of Landsat data respectively. NDVI, NDSI, NDBI, NBR, and EVI are abbreviations for the normalized difference vegetation index, the normalized difference snow index, the normalized difference built-up index, the normalized burn ratio, and the enhanced vegetation index, respectively.

Parameter				Set	tings					
	1	2	3	4	5	6	7	8	9	10
maxSegments	6	8	10	8	10	6	8	8	8	8
spikeThreshold	0.9	0.9	0.9	0.5	0.5	0.9	0.9	0.9	0.9	0.9
preventOneYearRecovery	TRUE	TRUE	TRUE	TRUE	TRUE	FALSE	TRUE	TRUE	TRUE	TRUE
recoveryThreshold	0.25	0.25	0.25	0.25	0.25	0.25	0.5	0.75	0.25	0.25
pvalThreshold	0.05	0.05	0.05	0.05	0.05	0.05	0.05	0.05	0.1	0.05
bestModelProportion	0.75	0.75	0.75	0.75	0.75	0.75	0.75	0.75	0.75	0.5

Table S3. Ten settings of LandTrendr parameters tested in this study.

Noted other parameters are set as default as those provided in <u>https://emapr.github.io/LT-GEE/lt-gee-requirements.html</u>.

Table S4. Annual cropland classification re	esults for each agricultural zone under
different LandTrendr parameter settings.	

Agricultural zone	Parameter settings with	Statistics of F1 scores		
	the highest F1 score			
		Highest	Mean	Std
Huang-Huai-Hai Plain	3	0.89	0.87	0.01
Loess Plateau	4	0.83	0.80	0.04
Middle-lower Yangtze Plain	5	0.80	0.80	0.02
Northeast China Plain	4	0.88	0.87	0.01
Northern arid and semiarid region	6	0.71	0.62	0.06
Qinghai Tibet Plateau	8	0.67	0.59	0.10
Sichuan Basin and surrounding regions	5	0.78	0.76	0.02
Southern China	4	0.69	0.65	0.06
Yunnan-Guizhou Plateau	5	0.75	0.71	0.04

Std: standard deviations.

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Year	F1	OA	UA	PA	Kappa
1986	0.76	0.91	0.77	0.75	0.70
1987	0.76	0.92	0.78	0.75	0.71
1988	0.77	0.92	0.78	0.75	0.71
1989	0.77	0.92	0.78	0.76	0.72
1990	0.76	0.92	0.78	0.75	0.71
1991	0.77	0.92	0.78	0.76	0.72
1992	0.78	0.92	0.78	0.77	0.73
1993	0.77	0.92	0.78	0.77	0.72
1994	0.78	0.92	0.78	0.77	0.73
1995	0.78	0.92	0.79	0.77	0.73
1996	0.78	0.92	0.79	0.77	0.73
1997	0.78	0.92	0.79	0.77	0.73
1998	0.78	0.92	0.79	0.78	0.74
1999	0.79	0.92	0.79	0.79	0.74
2000	0.79	0.93	0.79	0.80	0.75
2001	0.80	0.93	0.79	0.80	0.75
2002	0.79	0.92	0.79	0.80	0.75
2003	0.79	0.92	0.79	0.79	0.74
2004	0.79	0.93	0.79	0.80	0.75
2005	0.80	0.93	0.80	0.80	0.76
2006	0.80	0.93	0.80	0.81	0.76
2007	0.80	0.93	0.80	0.80	0.76
2008	0.80	0.93	0.80	0.81	0.76
2009	0.81	0.93	0.81	0.81	0.77
2010	0.81	0.93	0.80	0.82	0.77
2011	0.82	0.93	0.81	0.82	0.78
2012	0.82	0.94	0.81	0.83	0.78
2013	0.82	0.94	0.81	0.82	0.78
2014	0.82	0.94	0.82	0.82	0.78
2015	0.82	0.94	0.82	0.83	0.79
2016	0.81	0.93	0.80	0.82	0.77
2017	0.81	0.93	0.81	0.81	0.77
2018	0.81	0.93	0.81	0.81	0.77
2019	0.80	0.93	0.80	0.80	0.76
2020	0.80	0.93	0.80	0.80	0.76
2021	0.80	0.93	0.79	0.81	0.76
Mean	0.79 ± 0.02	0.93 ± 0.01	0.79 ± 0.01	0.79 ± 0.02	0.75 ± 0.02

Table. S5. Pixel-wise accuracy of CACD calculated based on the annual validation samples. F1: F1 score. OA: overall accuracy. Kappa: Kappa coefficient. UA: user's accuracy. PA: producer's accuracy.

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Tolerance (years)	Accuracy
0	0.76
± 1	0.79
± 2	0.81
± 3	0.84
± 4	0.86
± 5	0.87

 Table. S6. Accuracy of CACD for the year of change under different tolerance years.

References

Crist, E.P. (1985). A TM Tasseled Cap equivalent transformation for reflectance factor data. *Remote Sensing of Environment*, *17*, 301-306