

## Supplementary Information

**Table S1.** The allometric equation of trees for estimating field AVWS.

Tree Species	Position	Empirical formula	Quote
Cypress	Branch	$W_b=0.0350*(D^2H)^{(0.7119)}$	[1]
	Stem	$W_s=0.0754*(D^2H)^{(0.7934)}$	
	Leaf	$W_l=0.0685*(D^2H)^{(0.6583)}$	
Locust	Branch	$W_b=58.6653*(D^2H)^{(0.18258)}$	[5]
	Stem	$W_s=207.6566*(D^2H)^{(0.8112)}$	
	Leaf	$W_l=11.7388*(D^2H)^{(0.6406)}$	
Chinese pine	Branch	$W_b=0.0689*D^{2.1435}$	[2]
	Stem	$W_s=0.0967*D^{2.126}$	
	Leaf	$W_l=0.0133*D^{2.4283}$	
Fir	Branch	$W_b=0.0037*D^{(2.7386)}$	[3]
	Stem	$W_s=0.0405*D^{(2.568)}$	
	Leaf	$W_l=0.0014*D^{(2.9302)}$ , $D<40$ cm $W_l=29.541*\ln(D)-63.15$ , $D>40$ cm	
Picea-Abies	Branch	$W_{branch}=0.0037*D^{2.7386}$	[3]
	Stem	$W_{steam}=0.0450*D^{205860}$	
	Leaf	$W_{leaf}=0.0014*D^{2.9302}$	
Poplar	Branch	$W_b=0.3521*D^{(1.5212)}$	[5]
	Stem	$W_s=0.1745*D^{(2.1667)}$	
	Leaf	$W_l=0.0008*D^{(2.9066)}$	
Oak tree	Branch	$W_b=0.0261*D^{(2.4658)}$	[5]

	Stem	$W_s=0.0346 \cdot D^{(2.6758)}$	
	Leaf	$W_l=0.0004 \cdot D^{(2.9285)}$	
Birch	Branch	$W_{branch}=12.853 \cdot D^{0.3021}$	[4]
	Stem	$W_{steam}=22.426 \cdot D^{0.385}$	
	Leaf	$W_{leaf}=5.3765 \cdot D^{0.3916}$	
Other broad leaved tree species	Branch	$W_b=0.5261 \cdot (D)^{(1.5945)}$	[5]
	Stem	$W_s=0.0147 \cdot (D)^{(2.8094)}$	
	Leaf	$W_l=0.0987 \cdot (D)^{(1.4916)}$	

**Table S2.** Sentinel-2 Satellite sensor parameters

Sensor	Band number	Band name	Wavelength (nm)	Resolution (meters)
MSI	1	Coastal aerosol	433-453	60
MSI	2	Blue	458-523	10
MSI	3	Green	543-578	10
MSI	4	Red	650-680	10
MSI	5	Vegetation Red Edge	698-713	20
MSI	6	Vegetation Red Edge	733-748	20
MSI	7	Vegetation Red Edge	773-793	20
MSI	8	NIR	785-900	10
MSI	8a	Narrow NIR	855-875	20
MSI	9	Water vapour	935-955	60
MSI	10	SWIR – Cirrus	1360-1390	60
MSI	11	SWIR-1	1565-1655	20
MSI	12	SWIR-2	2100-2280	20

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**Table S3.** Landsat 8 Satellite sensor parameters

Sensor	Band number	Band name	Wavelength(nm)	Resolution (meters)
OLI	Band1	Coastal aerosol	430-450	30
	Band2	Blue	450-515	30
	Band3	Green	525-600	30
	Band4	Red	630-680	30
	Band5	NIR	845-885	30
	Band6	SWIR1	1560-1660	30
	Band7	SWIR2	2100-2300	30
	Band8	PAN	503-676	15
	Band9	Cirrus	1360-1390	30
TIRS	Band10	TIRS1	10600-11210	100
	Band11	TIRS2	11500-12500	100

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**Table S4.** The statistics of the land cover in Jiuzhaigou National Nature Reserve

Type	Pixels (number)	Total area (ha)	Percentage
Snow	1763954	17639.54	27.12%
Water Body	9756	97.56	0.15%
Bare Soil	1171417	11714.17	18.01%
Coniferous Forest	1866762	18667.62	28.70%
Broad-leaved Forest	1036300	10363.00	15.93%
Mixed Forests	109295	1092.95	1.68%
Shrubwood	166204	1662.04	2.56%
Grassland	380568	3805.68	5.85%
Total	6504256	65042.56	1.00

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**Table S5.** Vegetation indices to estimate AVWS.

Number	Vegetation Index Types	Calculation Formula	References
1	Perpendicular Vegetation Index	$PVI = \frac{R_{nir} - \alpha R_{red} - \beta}{\sqrt{\alpha^2 + 1}}$	[6]
2	Triangle Vegetation Index	$TVI = 0.5 \cdot [120 \cdot (R_{nir} - R_{green}) - 200 \cdot (R_{red} - R_{green})]$	[7]
3	Difference Vegetation Index	$DVI = R_{nir} - R_{red}$	[8]
4	Normalized Green-Blue Difference Index	$NGBDI = \frac{R_{green} - R_{blue}}{R_{green} + R_{blue}}$	[9]
5	Green-Blue Ratio Index	$GBRI = \frac{R_{green}}{R_{blue}}$	[10]
6	Atmospherically Resistant Vegetation Index	$ARVI = \frac{(R_{560} - R_{660})}{(R_{560} + R_{660} + R_{480})}$	[11]
7	Green-Red Ratio Index	$GRI = \frac{R_{green}}{R_{red}}$	[12]
8	Specific Leaf Area Vegetation Index	$SLAVI = \frac{R_{nir}}{R_{red} + R_{mir}}$	[13]
9	Normalized Different Moisture Index	$NDMI = \frac{(R_{nir} - R_{swir1})}{(R_{nir} + R_{swir1})}$	[14]
10	Normalized Difference Infrared Index	$NDII = \frac{(R_{820} - R_{1600})}{(R_{820} + R_{1600})}$	[15]

**Notes:** Red, blue, green, NIR and Mir in the table represent the atmospheric corrected spectral reflectance in red band, blue band, green band, near infrared band and mid infrared band respectively.

**Table S6.** Texture indices to estimate AVWS.

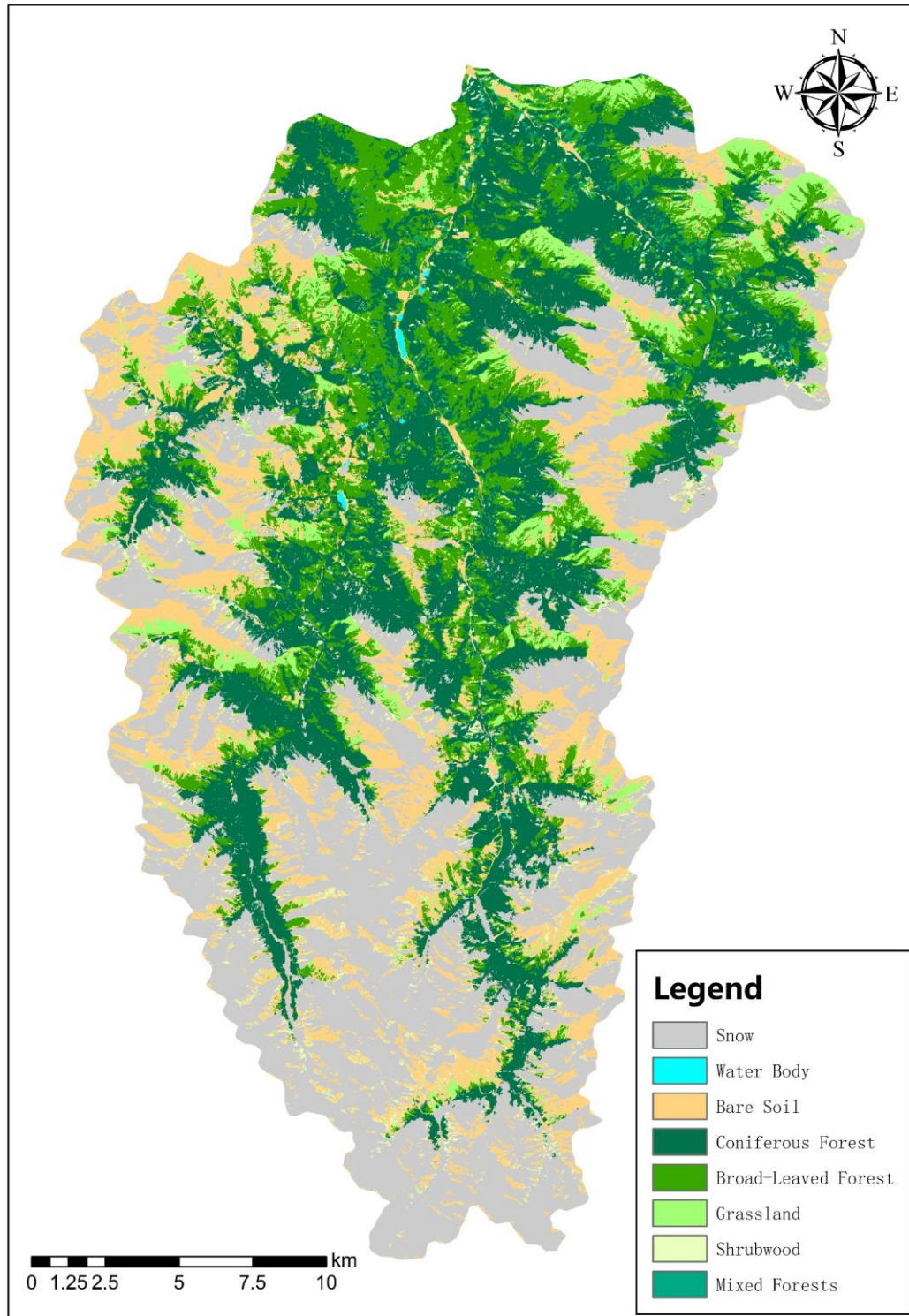
Serial Number	Texture Features	Calculation Formula
1	Mean	$\sum_{i,j=0}^{n-1} i(P_{i,j})$
2	Variance	$\sum_{i,j=0}^{n-1} iP_{i,j}(i - Mean)^2$
3	Homogeneity	$\sum_{i,j=0}^{n-1} i \frac{P_{i,j}}{1 + (i - j)^2}$
4	Contrast	$\sum_{i,j=0}^{n-1} iP_{i,j}(i - 1)^2$
5	Dissimilarity	$\sum_{i,j=0}^{n-1} iP_{i,j} (i - j) $
6	Entropy	$\sum_{i,j=0}^{n-1} iP_{i,j}(-\ln P_{i,j})$
7	Second Moment	$\sum_{i,j=0}^{n-1} iP_{i,j}^2$
8	Correlation	$\sum_{i,j=0}^{n-1} iP_{i,j} \left[ \frac{(i - Mean)(j - Mean)}{\sqrt{Variance_i \cdot Variance_j}} \right]$

**Notes:** i and j represent the row number and column number in the gray-level co-occurrence matrix, respectively.  $P_{i,j}$  represents the pixel value of row i and column j in the gray-level co-occurrence matrix. n is the number of distinct gray levels in the quantized.

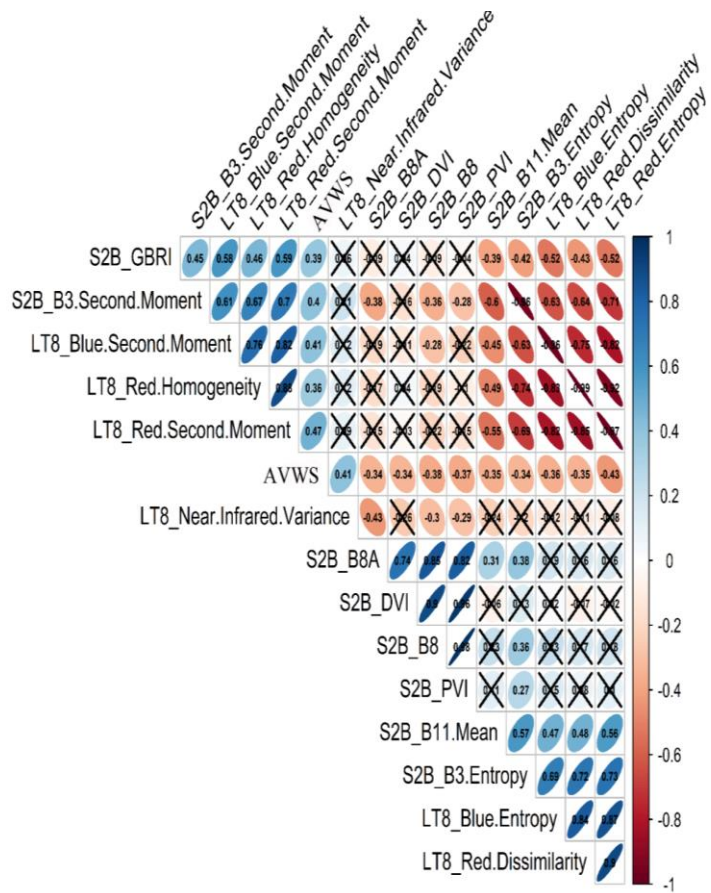
**Table S7.** The number of the variables input to the model.

Dataset	Model	Number of input features
Sentinel-2	MARS	5
	RF	5
	XGB	6
Landsat 8	MARS	5
	RF	4
	XGB	3
Sentinel-2+Landsat 8	MARS	7
	RF	6
	XGB	6

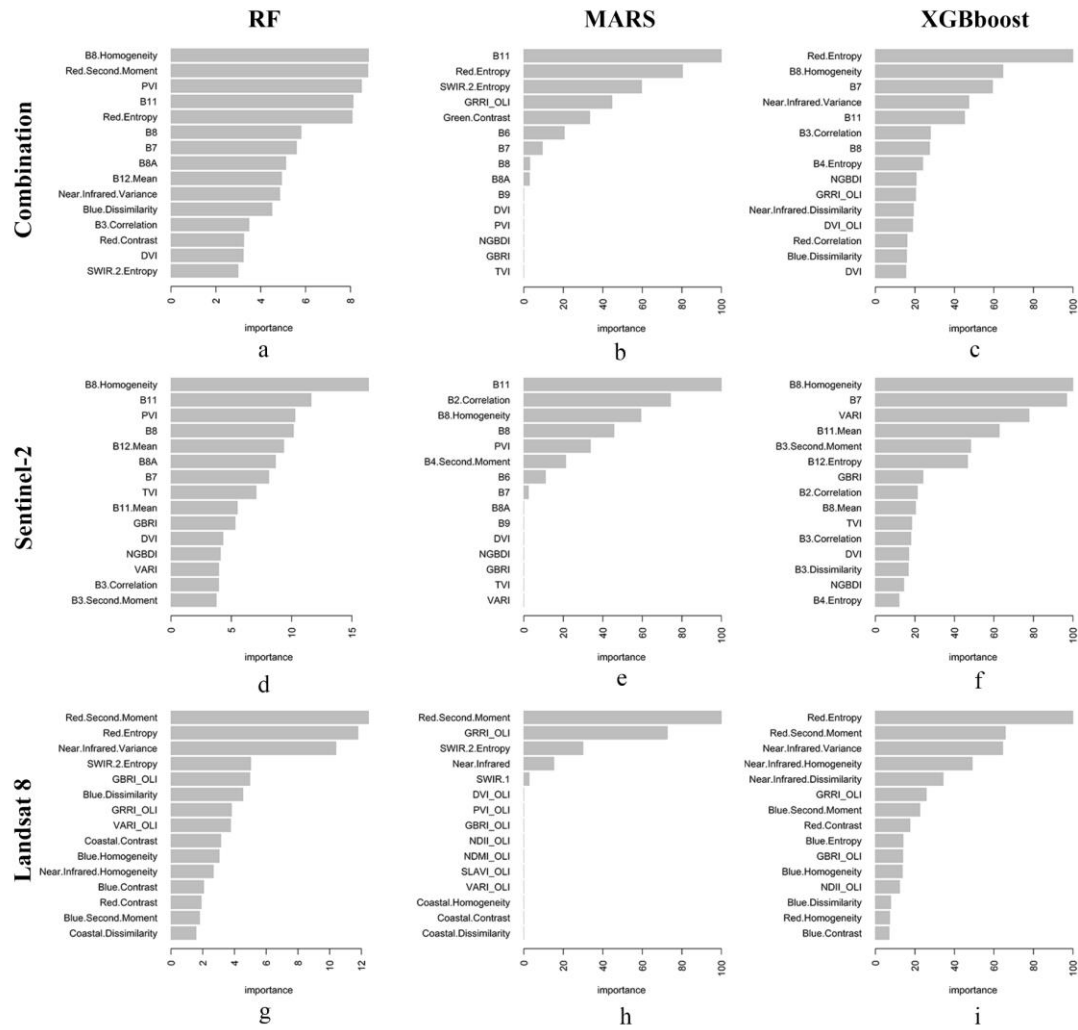




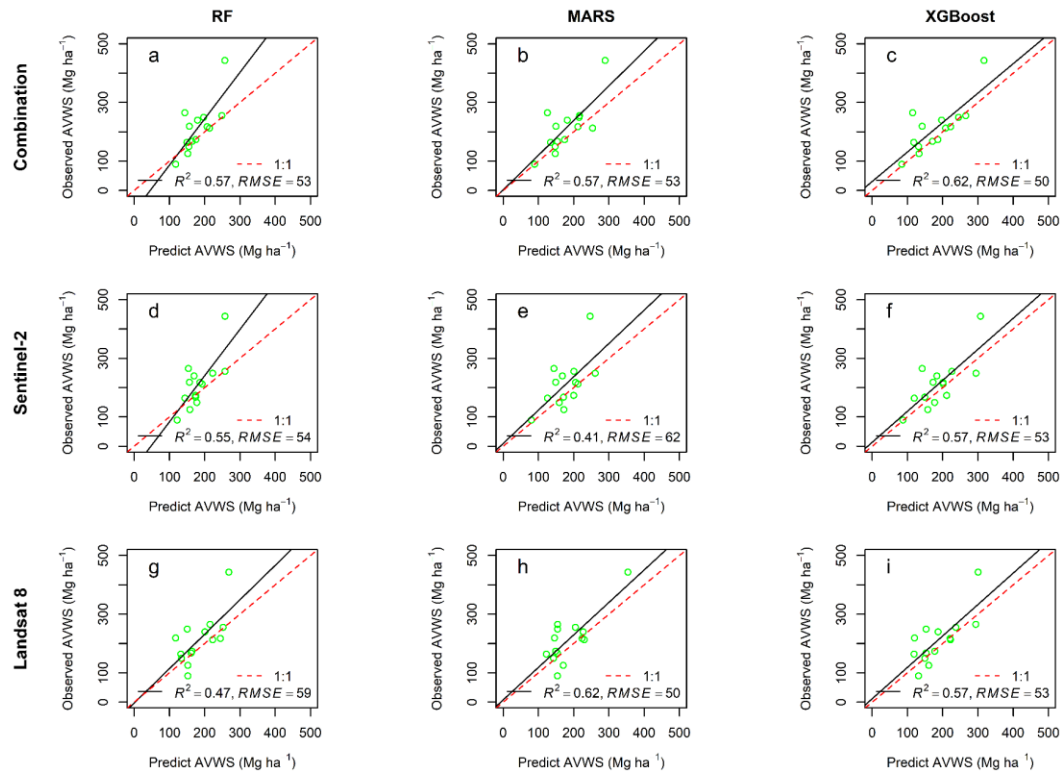
**Figure S1.** Classification of land cover in the study area



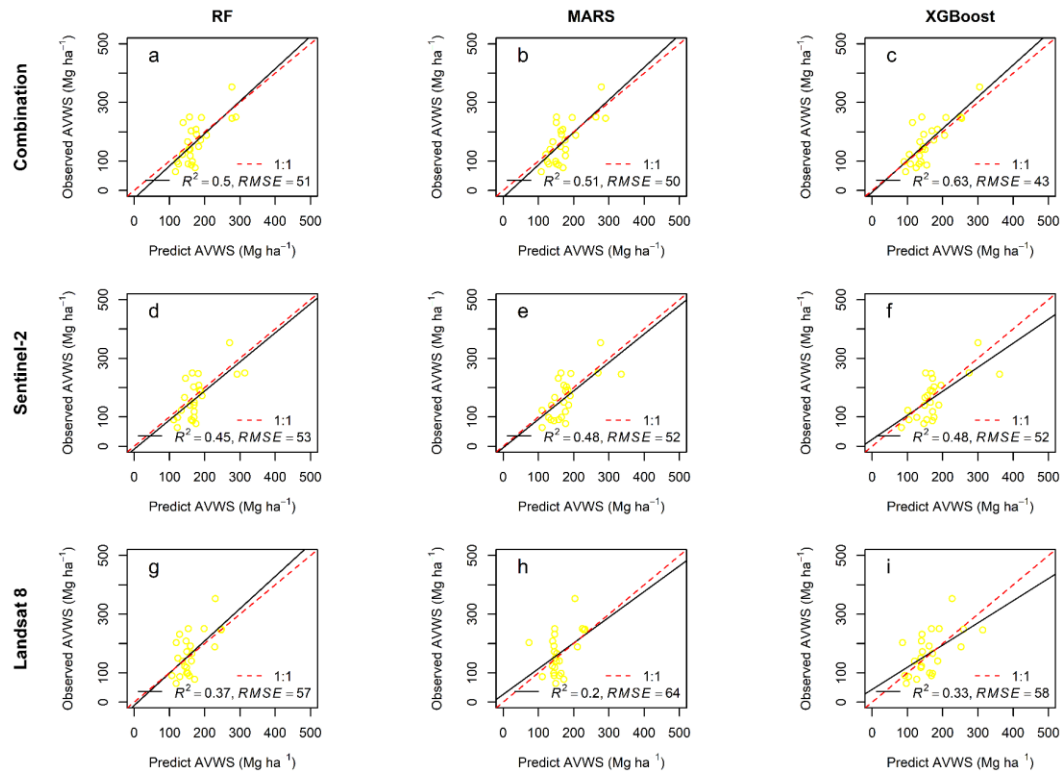
**Figure S2.** The correlations between AVWS and the variables of Sentinel-2B and Landsat 8 OLI. The 15 variables most correlated with AVWS were showed at a significance of  $p < 0.05$ .



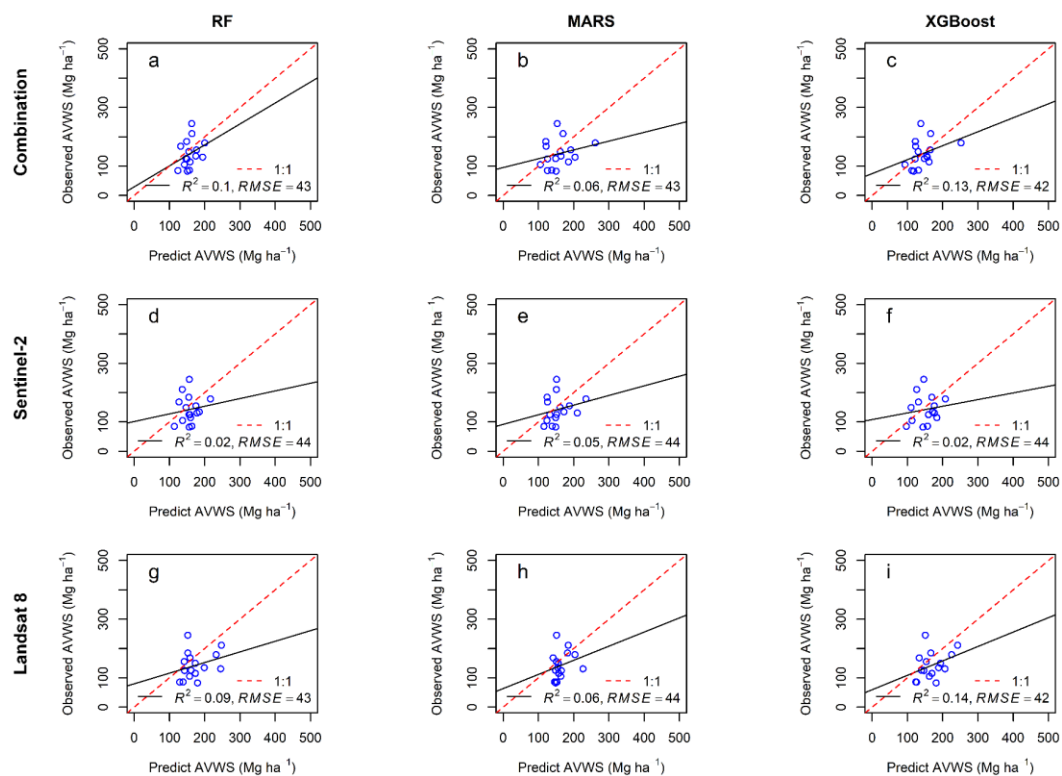
**Figure S3.** The Importance of variables from three dataset (Landsat 8, Sentinel-2 and their combination) using RF, XGB, and MARS algorithms.



**Figure S4.** The correlation between predicted and observed aboveground vegetation water storage (AVWS) of coniferous forest using different modelling approaches (XGBoost, RF and MARS) and satellite images (Landsat 8 and Sentinel-2).



**Figure S5.** The correlation between predicted and observed aboveground vegetation water storage (AVWS) of mixed forest using different modelling approaches (XGBoost, RF and MARS) and satellite images (Landsat 8 and Sentinel-2).



**Figure S6.** The correlation between predicted and observed aboveground vegetation water storage (AVWS) of broadleaved forest using different modelling approaches (XGBoost, RF and MARS) and satellite images (Landsat 8 and Sentinel-2).

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