



Forest Fire Detection Techniques Using Satellite Data

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1. Forest Fire Present Situation

2. Fire Detection using Satellite Images

3. Typical Application

1. Forest Fire Present Situation



➤ Forest Fire in abroad

More than **9250 km²** had been burned in August, 2019 in Amazon. There were **9507** times forest fire in one week (Aug. 22, 2019).



August 13, 2019 (from NASA)



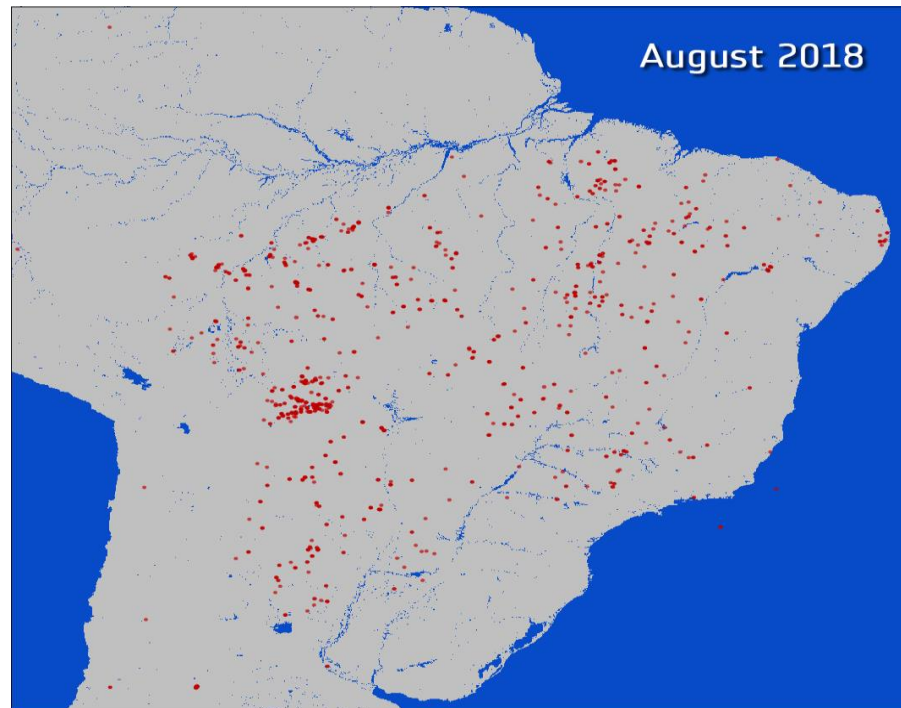
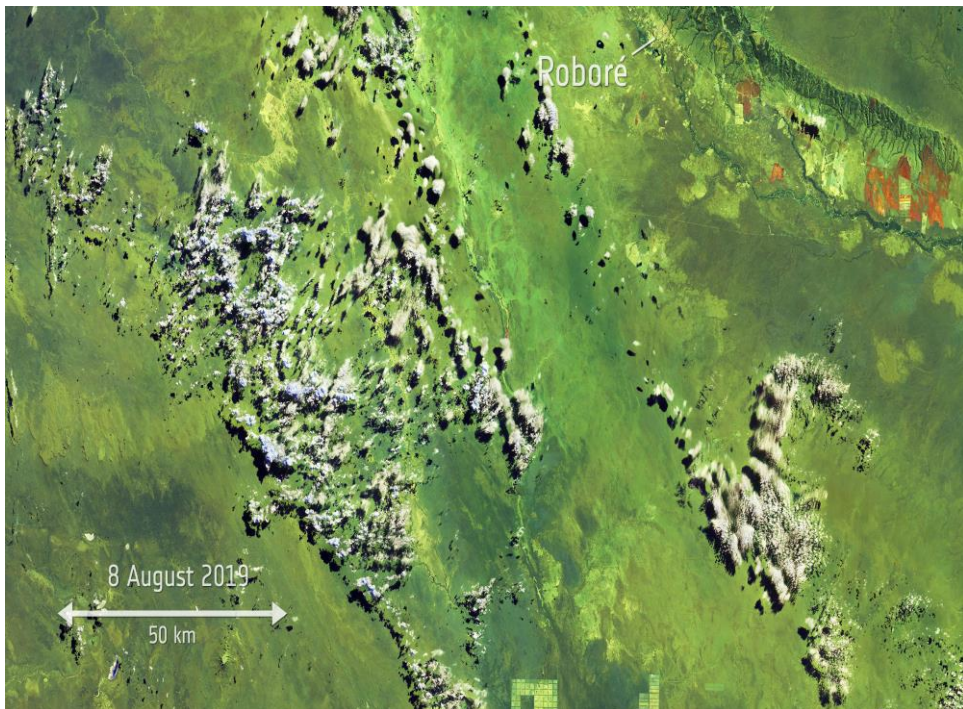
Sep. 15, 2019



1. Forest Fire Present Situation



➤ Forest Fire in abroad



BRASIL, August, 2019

(from ESA)



1. Forest Fire Present Situation



➤ Forest Fire in abroad



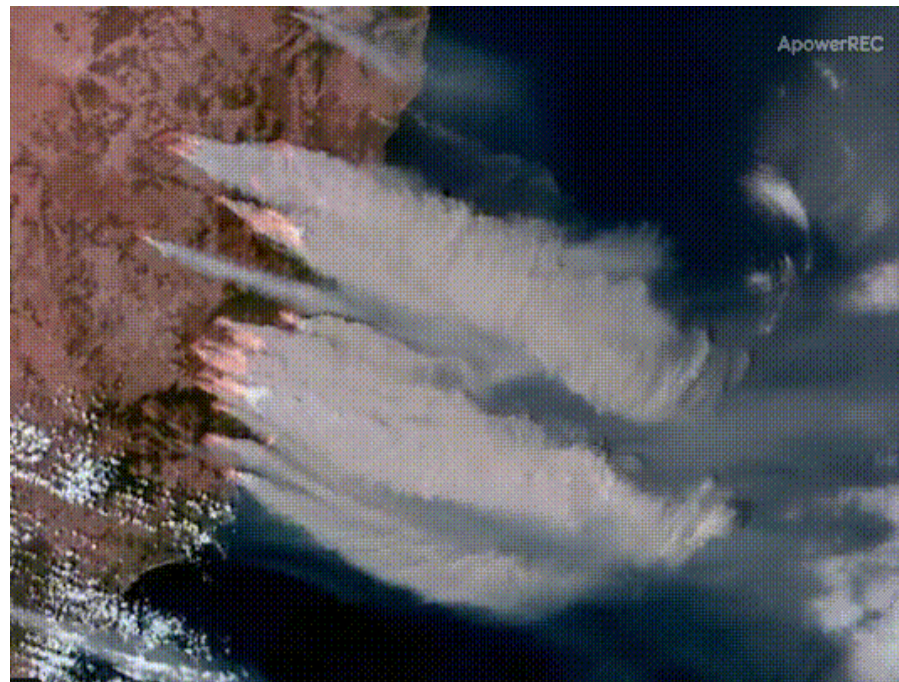
Australia, 2019.11.11



1. Forest Fire Present Situation



➤ Forest Fire in abroad



Australia, 2019.11.11



1. Forest Fire Present Situation



➤ Forest Fire in abroad



Australia, 2019.11.11



1. Forest Fire Present Situation



➤ Forest fire in China

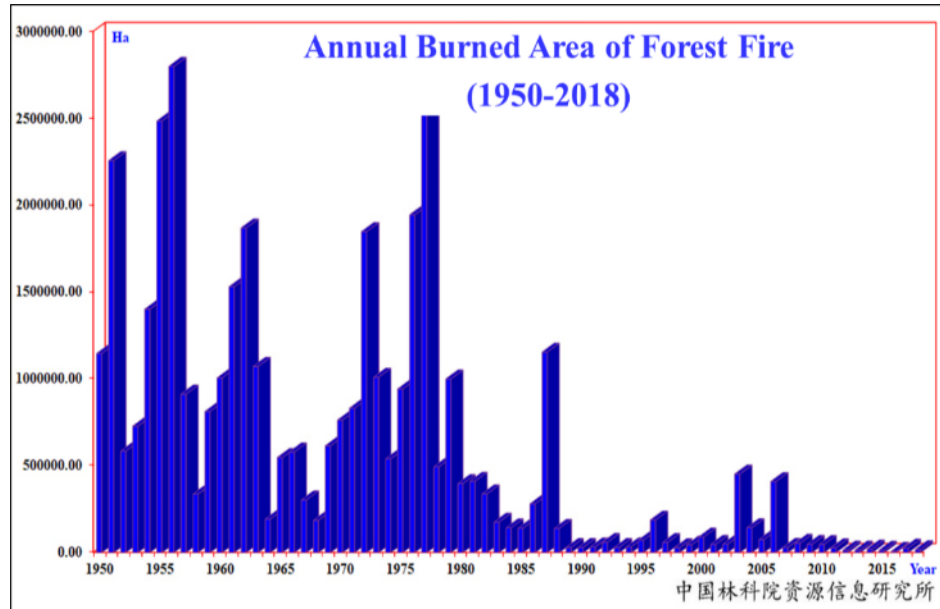
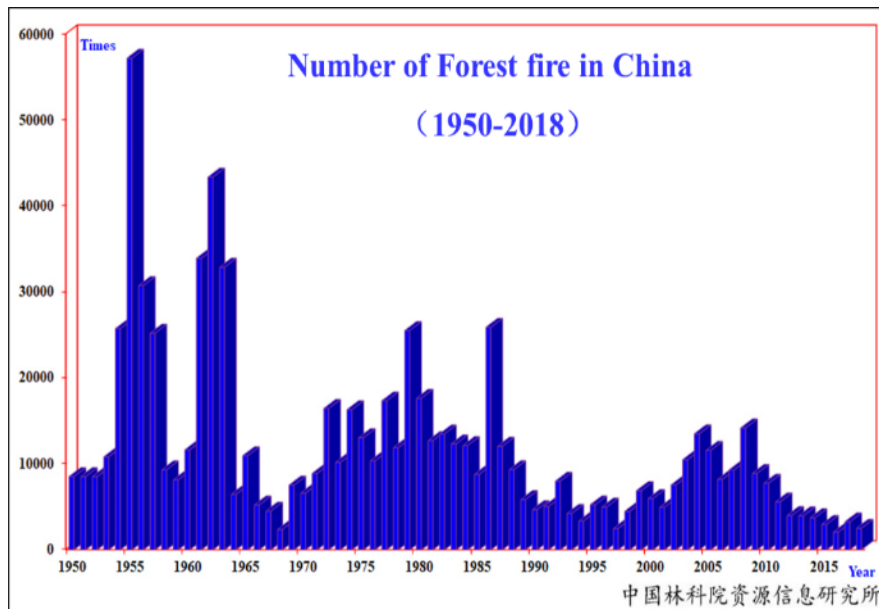


1. Forest Fire Present Situation



➤ Forest fire in China

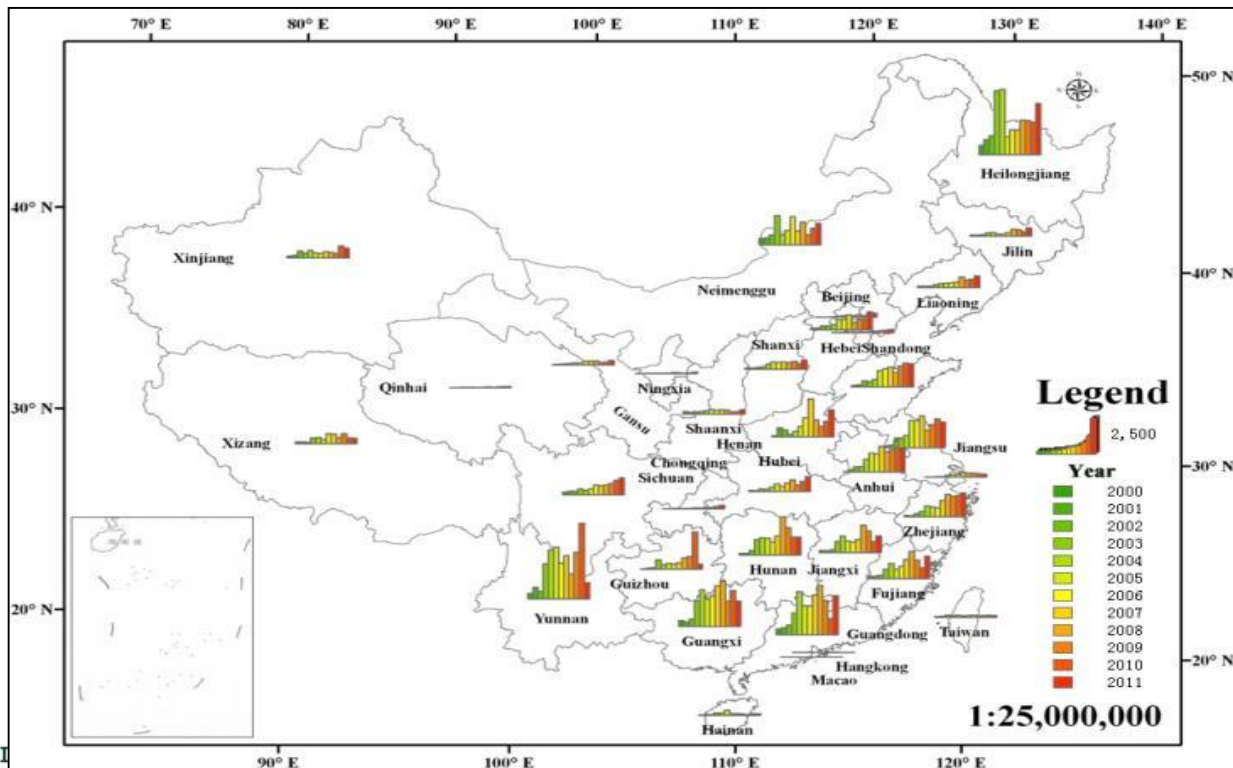
The annual average of forest fires is about 12,000 times and the damaged forest area is over 550,000 hm² in China from year 1950 to 2018.



1. Forest Fire Present Situation



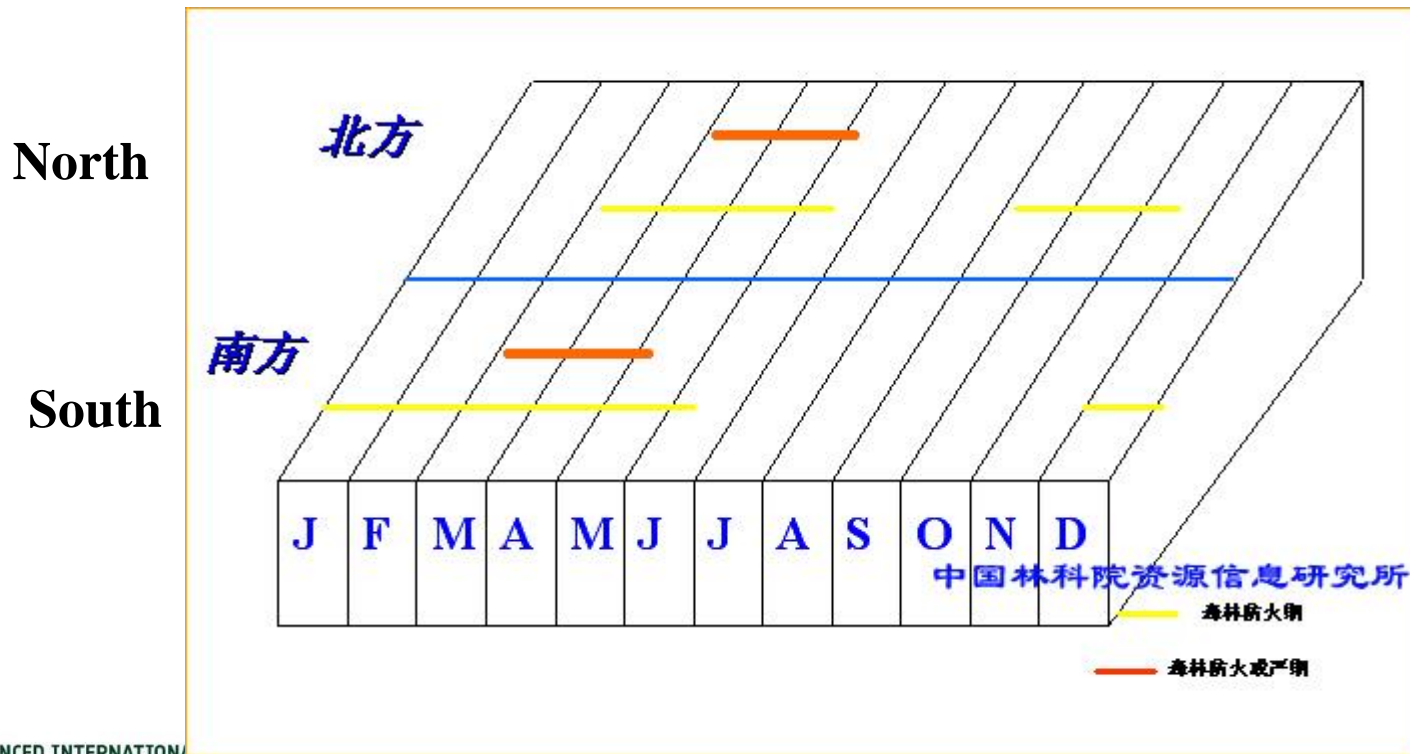
➤ Forest fire in China



1. Forest Fire Present Situation



➤ Forest fire season of China



1. Forest Fire Present Situation



➤ Regulations of Chinese forest fire prevention

Type	1988.1.16	2009.1.1
火警 (Fire alarm)	The damaged forest area is less than 1hm² or ignition in other forest land	/
一般森林火灾 (General forest fire)	The damaged forest area is between 1hm² and 100hm²	The damaged forest area is less than 1hm² or ignition in other forest land; or died person is between 1 and 3 ; or hurt person is between 1 and 10
较大森林火灾 (Large forest fire)	/	The damaged forest area is between 1hm² and 100hm² ; or died person is between 3 and 10 ; or hurt person is between 10 and 50
重大森林火灾 (Serious forest fire)	The damaged forest area is between 100hm² and 1000hm²	The damaged forest area is between 100hm² and 1000hm² ; or died person is between 10 and 30 ; or hurt person is between 50 and 100
特别重大森林火灾 (Extra serious forest fire)	The damaged forest area is larger than 1000hm²	The damaged forest area is larger than 1000hm² ; or more than 30 person died; or more than 100 person hurt

我国的森林防火条例



森林火灾是林地上失控的火，它是自由蔓延、超过一定面积、造成一定损失的林火。

类型	1988年1月16日颁布	2009年1月1日起执行
火警 (Fire alarm)	受害森林面积 不足1公顷 或者其他林地起火的	/
一般森林火灾 (General forest fire)	受害森林面积在 1公顷 以上 不足100公顷 的	受害森林面积在 1公顷 以下或者其他林地起火的，或者 死亡1人 以上 3人 以下的，或者重伤 1人 以上 10人 以下的
较大森林火灾 (Large forest fire)	/	受害森林面积在 1公顷 以上 100公顷 以下的，或者 死亡3人 以上 10人 以下的，或者重伤 10人 以上 50人 以下的
重大森林火灾 (Serious forest fire)	受害森林面积在 100公顷 以上 不足1000公顷 的	受害森林面积在 100公顷 以上 1000公顷 以下的，或者 死亡10人 以上 30人 以下的，或者重伤 50人 以上 100人 以下的
特别重大森林火灾 (Extra serious forest fire)	受害森林面积在 1000公 顷 以上的	受害森林面积在 1000公顷 以上的，或者 死亡30人 以上的，或者重伤 100人 以上的

1. Forest Fire Present Situation



➤ Fire Classification

- Forest Fire
- Grassland fire
- Prescribed burning (Fire in Silviculture)
- Agricultural fire (prescribed residues burning, crop residue burning, etc.)



2. Fire Detection



➤ The Four Steps of Forest Fire-fighting

● Prevention

Previous works of cleanliness of the forest: tracing firewalls, preparation of firefighting strategies...

● Warning

Fire Risk Indexes Maps: **fuel and vegetation parameters monitoring (TIR)**, meteorological conditions, human factor ...

● Crisis

Fire detection and monitoring (SWIR/MIR/TIR); fire propagation simulator.

● Post-crisis

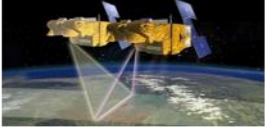
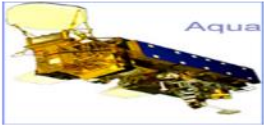




Mid and long-term damage assessment and vegetation evolution.



2. Fire Detection



➤ Fire Monitoring System

Satellite	NOAA/AVHRR 	Terra/Aqua 	MODIS	FY 	Space
Airplane		UAV		Air	
Fire Watch		Video image		Near Ground	
Watch man				Ground	

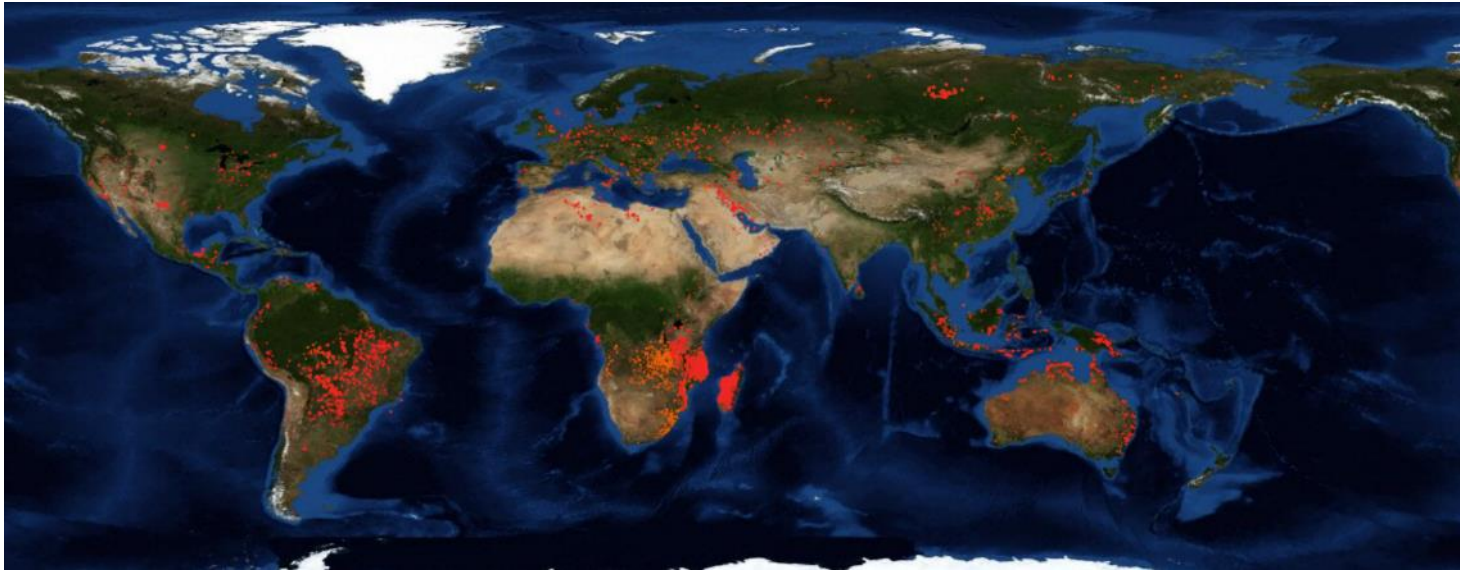


2. Fire Detection



➤ Active Fire Monitoring using Satellite Images

Satellite remote sensing has been a unique technique to monitor global active fire.



(<https://firms.modaps.eosdis.nasa.gov/map>, August 25, 2019)



2. Fire Detection



➤ Principle of hotspots identification

● Planck's Law

$$M_{\lambda} = \pi L_{\lambda} = \frac{\pi 2hc^2 \lambda^{-5}}{\exp\left(\frac{hc}{k\lambda T}\right) - 1}$$

$$M_{\lambda} = \frac{\pi k_1 \lambda^{-5}}{\exp\left(\frac{k_2}{\lambda T}\right) - 1}$$

$$c = 3.0 \times 10^8 \text{ m/s}$$

$$k = 1.38 \times 10^{-23} \text{ J.K}^{-1}$$

$$h = 6.62 \times 10^{-34} \text{ J.s}$$

$$k_1 = 3.74 \times 10^{-16} \text{ W.m}^2$$

$$k_2 = 1.44 \times 10^{-2} \text{ m.K}$$



● Stefan-Boltzman's Law

$$M = \int_0^{\infty} M_{\lambda} d\lambda = \frac{2\pi^5 k^4 T^4}{15c^2 h^3} = \sigma T^4$$

$$\sigma = 5.67 \times 10^{-8} \text{ W.m}^{-2}.\text{K}^{-4}$$

The total emittance from a black body, including all wavelengths, is directly proportional to the fourth power of its temperature.

This temperature is call “Brightness Temperature, BT”, and it is what we directly obtain from the satellite signal.

● Wien's Law

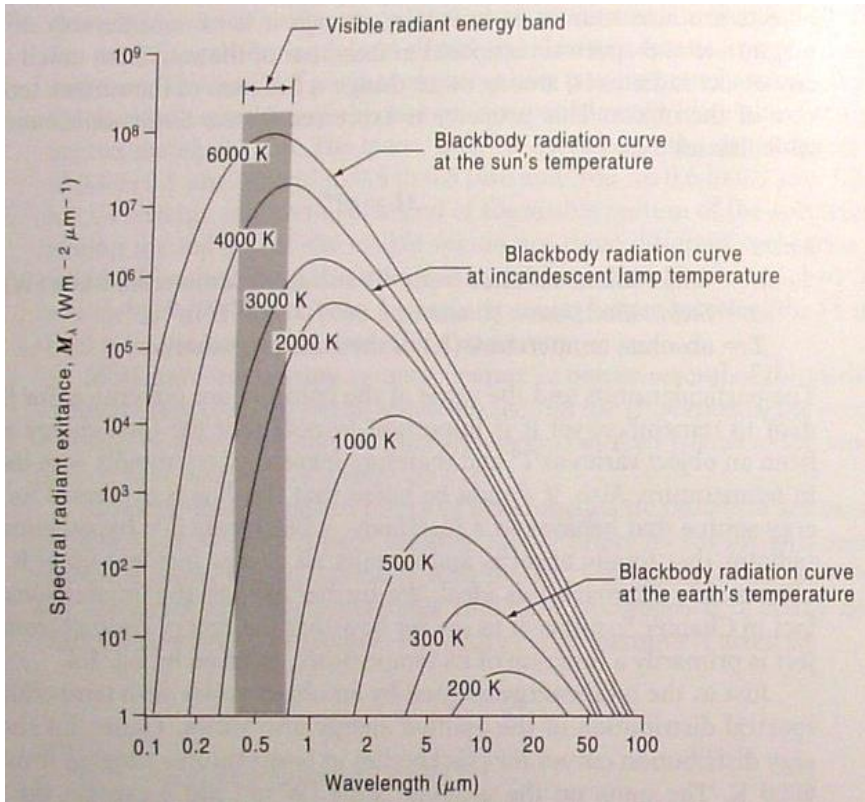
$$\lambda_{\max} = p_1/T$$

$$M_{\lambda_{\max}} = p_2 \cdot T^5$$

$$p_1 = 2.898 \times 10^{-3} \text{ m.K}, p_2 = 1.2862 \times 10^{-5}$$

The maximum **spectral emittance** is proportional to the **fifth** power of the **temperature**, and it corresponds to a peak wavelength, λ_{\max} , which is **inversely proportional** to the **temperature**.

➤ Principle of Hotspots Identification



The λ_{MAX} Displacement from the Law of Wien.

Examples of Wien's Law

$$T (\text{sun}) = 6000K$$

$$\lambda_{max} = k/T = 2898/6000 = 0.483\mu m$$

$$T (\text{earth}) = 300K$$

$$\lambda_{max} = k/T = 2898/300 = 9.66\mu m$$



● Emissivity

The real bodies are not black bodies, BB's. Its emittance is always lower than that of the black bodies. So we need to introduce a factor called "emissivity", ε , being $\varepsilon < 1$, to obtain its emittance. Then, they are called "grey bodies".

$$M_{GB} = \varepsilon \cdot \sigma \cdot T_{GB}^4; \quad \varepsilon < 1$$

We can imagine a BB with the same emittance as the GB

$$M_{BB} = M_{GB} = \sigma \cdot T_{BB}^4$$

To have the same emittance as the grey body, this BB has to be at a temperature T_{BB}

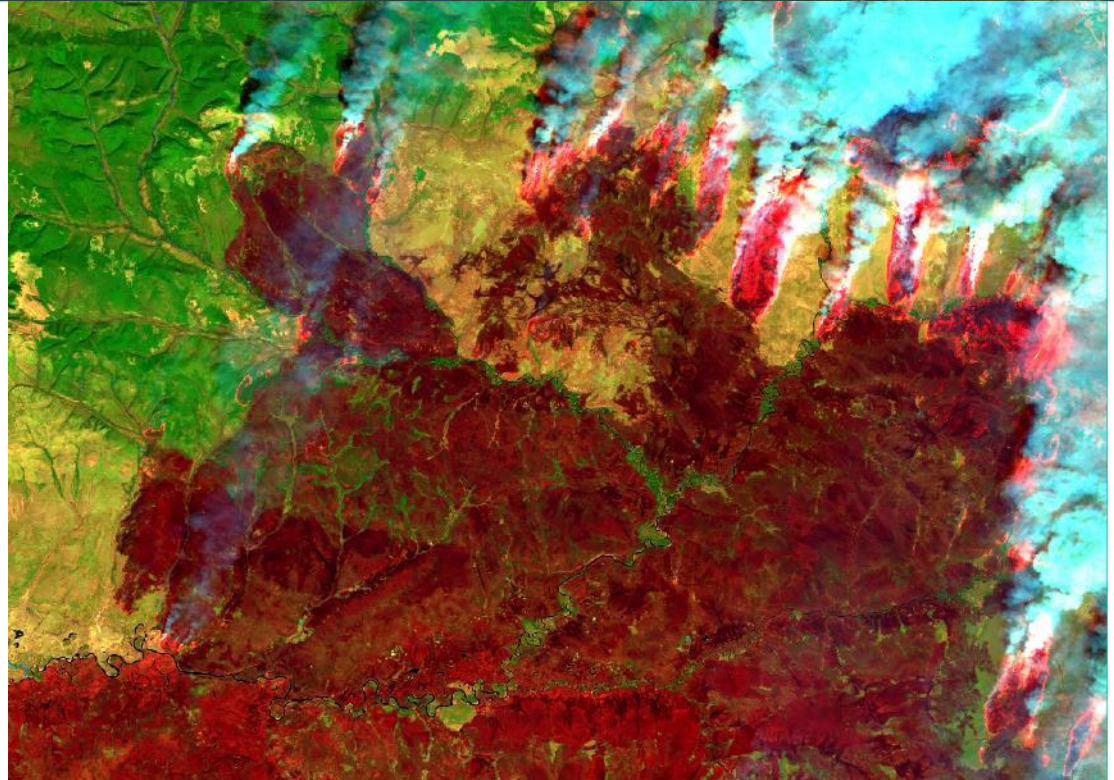
$$T_{BB} = \varepsilon^{1/4} \cdot T_{GB}$$

2. Fire Detection



➤ Combustion status

- Smoking
- Flaming
- Smouldering
- Burned area



S2B, 2018.5.9



2. Fire Detection



➤ Purpose

- **Detect burning in the vast forest areas**
- **Map large fires to support suppression**
- **Monitor all phases of a fire in the key regions**



2. Fire Detection



➤ Satellite Data

Low Spatial Resolution Optical Data

- NOAA-AVHRR (1100m)
- MODIS (1000m)
- ENVISAT-AATSR (1100m)
- Himawari-8/9 (500/2000m)
- VIIRS (375/750m)
- **Sentinel-3 SLSTR (1000m)**
- **FY Series (1100/1000m)**

Mid-high Spatial Resolution optical Data

- TM/ETM⁺/Landsat 8 OLI (30m)
- SPOT (10/20m)
- **Sentinel-2 (10/20/60m)**
- HJ-1B IRS (30/150/300m)
- **GF-4 (50/400m)**



Comparison of the Sentinel-3 SLSTR, AATSR and ATSR-1/2 instruments



	Capability	SLSTR	AATSR & ATSR-1 & 2
Swath	Nadir view	>1400 km	500 km
Dual view	>740 km	500 km	
Global coverage revisit times	1 satellite (dual view)	1.9 days (mean)	3 days at mid-latitudes
2 satellites (dual view)	0.9 day (mean)	–	
1 satellite (nadir view)	1 day (mean)	3 days at mid-latitudes	
2 satellites (nadir view)	0.5 day (mean)	–	
Spatial sampling interval at SSP (km)		0.5 km VIS-SWIR 1 km IR-Fire	1 km
Spectral channel Centre, λ (μm)	VIS SWIR MWIR/TIR Fire1/2	0.555; 0.659; 0.865 1.375; 1.610; 2.25 3.74; 10.85; 12 3.74; 10.85	0.555; 0.659; 0.865 ^a 1.610 3.74; 10.85; 12
Radiometric resolution	VIS (A = 0.5%) SWIR (A = 0.5%)	SNR >20 SNR >20	SNR >20 SNR >20
	MWIR (T = 270K) TIR (T = 270K) Fire-1 (<500K) Fire-2 (<400K)	NEAT < 80mK NEAT < 50mK NEAT < 1K NEAT < 0.5K	NEAT < 80mK NEAT < 50mK
Radiometric accuracy	VIS-SWIR (A = 2–100%)	<2% (BOL) <5% (EOL)	<5%
	MWIR-TIR (265–310K) Fire (<500K)	<0.2K (0.1K goal) <3K	<0.2K
Design lifetime ^b		7.5 years	ATSR-1 & 2: 3 years AATSR: 5 years

A, albedo; BOL, beginning of life; EOL, end of life; SSP, subsatellite point; NEAT, Noise-Equivalent Temperature Difference.

^a These channels were present for the AATSR and ATSR-2, but not the ATSR-1.

^b Some instruments remain in operation for much longer than their ‘design lifetimes’. Launched in 2002, Envisat’s AATSR, for example, was designed for 5 years, but continued to operate for almost 10 years until 2012. Similarly, ERS-1 has provided an uninterrupted series of ATSR-type data and data products since its launch in 1991.



The Sentinel-2 Mission



Band number	Central wavelength (nm)	Bandwidth (nm)	Spatial resolution (m)	L_{ref} ($W.m^{-2} sr^{-1} \mu m^{-1}$)	SNR @ L_{ref}
1	443	20	60	129	129
2	490	65	10	128	154
3	560	35	10	128	168
4	665	30	10	108	142
5	705	15	20	74.5	117
6	740	15	20	68	89
7	783	20	20	67	105
8	842	115	10	103	174
8b	865	20	20	52.5	72
9	945	20	60	9	114
10	1380	30	60	6	50
11	1610	90	20	4	100
12	2190	180	20	1.5	100



GF-4 Satellite



Have been launched in Xichang Satellite Launch Center on December 29, 2015. It's Geostationary orbit satellite. The fixed point location is 105.6° E.

To explore the forest fire monitoring method of GF-4 PMI images.

Sensor	Spectrum band No.	Spectral range (μm)	Spatial resolution (m)	Breadth (km)	Revisit time	Application for fire
Visible light near infrared (PMS)	1	0.45~0.90	50	500	20s	
	2	0.45~0.52				Smoke
	3	0.52~0.59				
	4	0.63~0.69				Burned area mapping
	5	0.77~0.89				
Medium wave infrared (IRS)	6	3.50~4.10	400	400		Fire identification



- **Image Processing**
(Image enhanced, RGB composition, PCA, etc.)
- **Classification**
(Supervised, Decision Tree, SVM, Random Forest Classification, etc.)
- **Threshold Value**
(Fixed threshold, Variable Threshold, Context threshold, etc.)
- **Vegetation Index**
(NDVI, Ratio Vegetation, etc.)
- **Artificial Intelligence**
(Neural Network , Expert System, Deep learning, etc.)

● Example

Algorithms based on threshold: These algorithms look for a minimum threshold of the **MIR** and **TIR** values and a minimum difference between **MIR** and **TIR** values. The **MIR** and **TIR** values are its brightness temperatures.

The general condition of this kind of algorithms is:

$$T_{\text{MIR}} > T_{\text{MIR, threshold}}$$

$$T_{\text{TIR}} > T_{\text{TIR, threshold}}$$

$$T_{\text{MIR}} - T_{\text{TIR}} > T_{\text{MT, threshold}}$$

Sometimes an additional condition is included:

$$\alpha_{\text{NIR}} < \alpha_{\text{threshold}}$$

Due to the fact that the burnt soil has a very low **NIR** reflectance.

● Example

ALGORITHMS BASED ON CONTEXTUAL ANALYSIS: These algorithms analyzes the **MIR** and **TIR** values on the analyzed pixel plus the pixels around it, and they check if the mean value and standard deviation of these values overcome some conditions. Then the core of this analysis is the variability of the pixels around the possible fire. The general condition of this kind of algorithms is:

$$\sigma_{\text{MIR}} > \sigma_{\text{MIR,threshold}}$$

$$\sigma(\text{T}_{\text{MIR}} - \text{T}_{\text{TIR}}) > \sigma_{\text{MT,threshold}}$$

Being σ the standard deviation.

Sometimes an additional condition is also included:

$$\alpha_{\text{NIR}} < \alpha_{\text{threshold}}$$

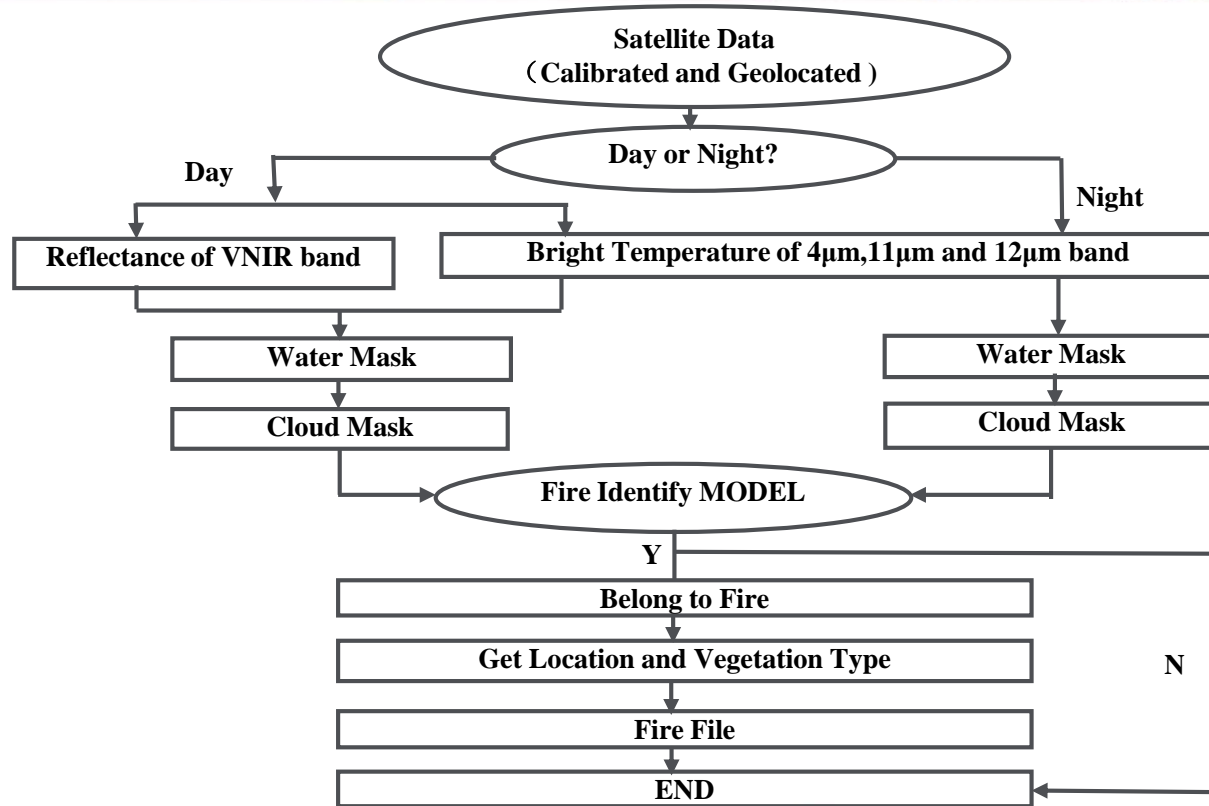
Due to the fact that the burnt soil has a very low **NIR** reflectance.

● Example

Table 2. Five fire detection algorithms tested over the Canadian boreal forest. Note that all statistics (av, ad, md, sd) refer to background pixels; all temperature values are expressed in degrees Kelvin (K); all thresholds given here refer to daytime data.

Algorithm / Description	CCRS	ESA	IGBP	GIGLIO	MODIS
Algorithm category and geographic applicability	fixed thresholds, regional (Canada)	fixed thresholds, global/regional	contextual, global	contextual, global	contextual, global
Potential fire detection			$T_3 > 311$ AND $T_{34} > 8$	$T_3 > 310$ AND $T_{34} > 6$	$T_3 \geq 315$ AND $T_{34} \geq 5$
Background window size			3x3 to 15x15	5x5 to 21x21	3x3 to 21x21
Min. number of pixels			Max {25% of pixels tested , 3}	Max {25% of pixels tested , 6}	Max {25% of pixels tested , 3}
Background selection			$T_3 \leq 311$ OR $T_{34} \leq 8$	$T_3 \leq 318$ OR $T_{34} \leq 12$	$T_3 \leq 320$ OR $T_{34} < 20$
Actual fire detection with T_3 and/or T_4	$T_3 > 315$	$T_3 > 320$	Define: $\xi_3 = \text{av}(T_3) + 2 * \text{sd}(T_3) + 3$ $\xi_{34} = \text{Max}\{8, \text{av}(T_{34}) + 2 * \text{sd}(T_{34})\}$ Then, Confirm potential fires as real if: $T_3 > \xi_3$ AND $T_{34} > \xi_{34}$	Define: $\xi_4 = \text{av}(T_4) + \text{ad}(T_4) - 3$ $\xi_{34} = \text{av}(T_{34}) + \text{Max}\{2.5 * \text{ad}(T_{34}), 4\}$ Then, Confirm potential fires as real if: $T_4 > \xi_4$ AND $T_{34} > \xi_{34}$	Define: $\xi_3 = \text{Min}\{320, \text{av}(T_3) + 4 * \text{Max}\{\text{sd}(T_3), 2\}\}$ $\xi_{34} = \text{Min}\{20, \text{md}(T_{34}) + 4 * \text{Max}\{\text{sd}(T_{34}), 2\}\}$] Then, Confirm potential fires as real if: $T_3 > 360$ OR $[T_3 > \xi_3 \text{ AND } T_{34} > \xi_{34}]$
Filter hot surfaces	$T_{34} \geq 14$	$T_{34} > 15$	Incorporated into fire detection	Incorporated into fire detection	Incorporated into fire detection
Filter clouds	$T_4 \geq 260$	$T_4 > 245$	$R_1 + R_2 \leq 1.2$ AND $T_2 \geq 265$ AND $(R_1 + R_2 \leq 0.8 \text{ OR } T_2 \geq 285)$	<i>IGBP criteria applied here (no external cloud mask).</i>	<i>IGBP criteria applied here (no external cloud mask).</i>
Filter reflective surfaces	$R_2 \leq 0.22$	$R_1 < 0.25$	$R_2 < 0.20$	$R_2 < 0.25$	
Filter sun glint		$ R_1 - R_2 > 0.01$			$R_1 \leq 0.3$ OR $R_2 \leq 0.3$ OR reflected sun angle $\geq 40^\circ$
Other detection criteria	$T_{34} \geq 19$ OR $T_{45} < 4.1$				
Post processing (not applied in this investigation)	<ul style="list-style-type: none"> Elimination of non-forest and isolated pixels 	<ul style="list-style-type: none"> Quicklook inspection, Max annual NDVI > 0 			

➤ Basic flowchart of Fire detection



Step 1: Pre-processing

(Radiometric Correction, Geometric Correction, etc.)

Step 2: Physical value calculation

(Reflectance, Bright Temperature, Vegetation Index, etc.)

Step 3: Characteristic Analysis (Sample)

Step 4: Constructing Model

Step 5: Validation Results

Step 6: Mapping Results

● Reflectance Calculation

$$\text{Radiance}(L_{\lambda}) = \text{scales} * \text{DN} - \text{offset}$$

$$\text{TOA_Ref} = \text{Radiance} * 1.0 / \cos(D2R * \theta_s)$$

Where: DN is the digital data

TOA_Ref is the Top-of-Atmosphere (TOA) reflectance

θ_s is solar zenith angle

D2R is degree to radian; the value is 0.0174533.

● Vegetation Index (VI)

$$\text{NDVI} = (\rho_{\text{nir}} - \rho_{\text{red}}) / (\rho_{\text{nir}} + \rho_{\text{red}})$$

● Radiance Calculation

$$\text{Radiance}(L_\lambda) = \text{scales} * \text{DN} - \text{offset}$$

● Bright Temperature Calculation

$$BT_\lambda = \frac{C_2}{\lambda * \ln(1.0 + \frac{C_1}{L_\lambda * \lambda^5})}$$

Where: BT_λ is the bright temperature of central wavelength;

λ is the central wavelength;(unit: μm);

L_λ is the radiance of Wavelength ; (unit: $\text{W micron}^4 / \text{m}^2 \text{ sr}$)

C_1, C_2 are constants;

Here, $C_1 = 1.19107\text{E}+8$ (unit: $\text{W micron}^4 / \text{m}^2 \text{ sr}$)

$C_2 = 1.43883\text{E}+4$ (unit: micron K)

● Bright Temperature Correction

$$T_{\lambda} = BT_{\lambda} * Ta_{\lambda} + Tc_{\lambda}$$

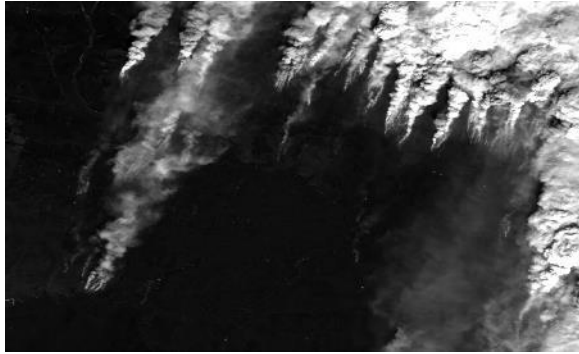
Where:

T_{λ} is the λ Wavelength's bright temperature after correction;

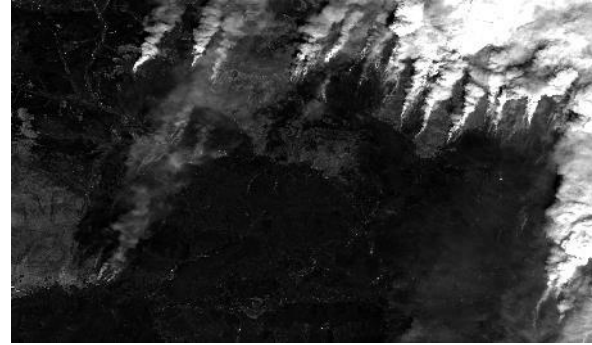
BT_{λ} Is the λ Wavelength's bright temperature before correction;

Ta_{λ} is the λ Wavelength's bright temperature correction scale;

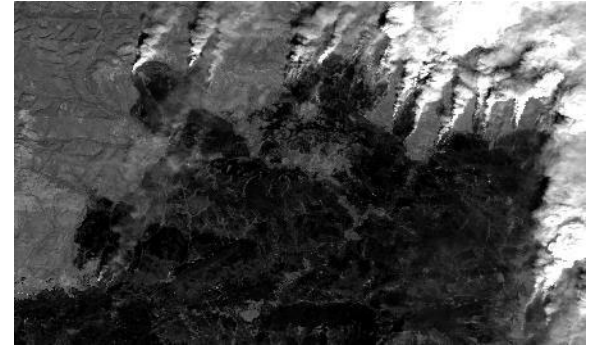
➤ Characteristics Analysis



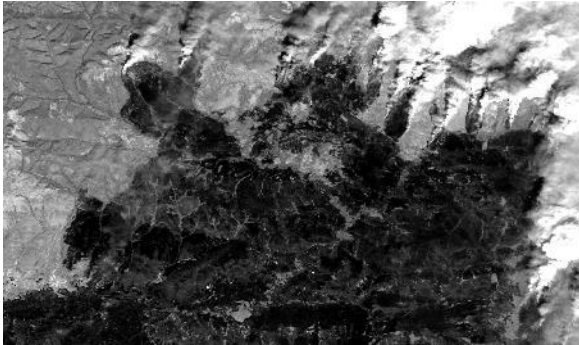
Blue(0.442μm)



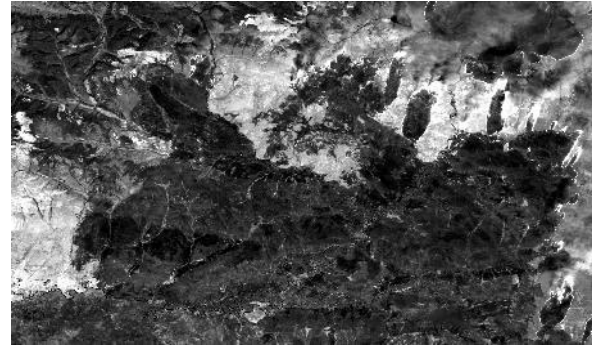
Red(0.665μm)



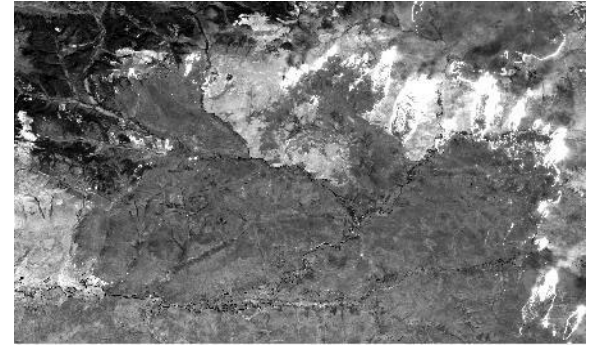
Red edge(0.739μm)



NIR(0.864μm)



SWIR(1.610μm)



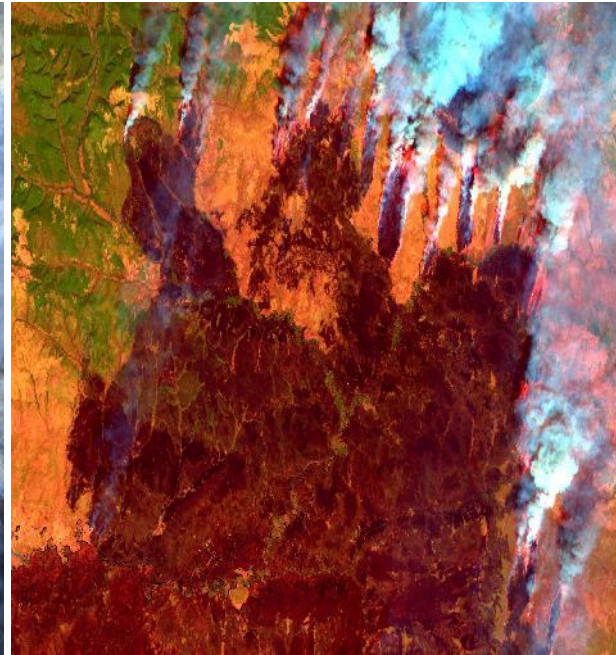
SWIR(2.186μm)



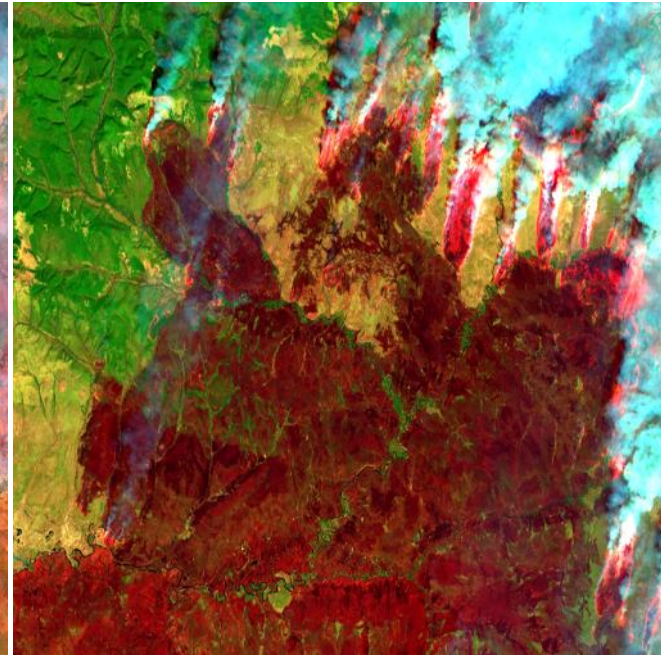
➤ Characteristics Analysis



RGB(0.665, 0.559, and 0.442 μ m)



RGB(1.610, 0.864, and 0.665 μ m)



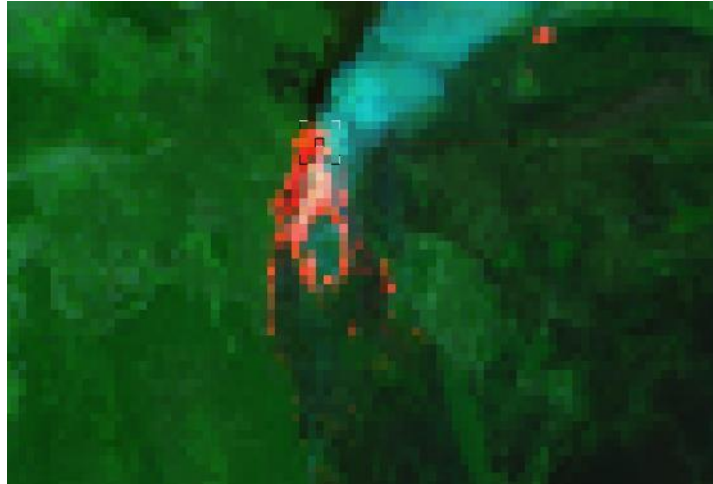
RGB(2.186, 0.864, and 0.665 μ m)



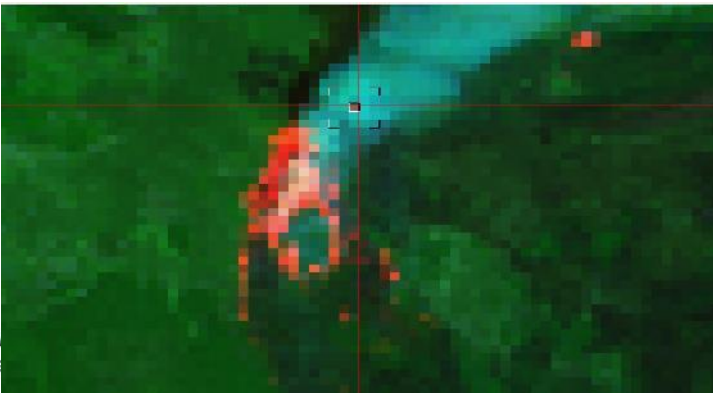
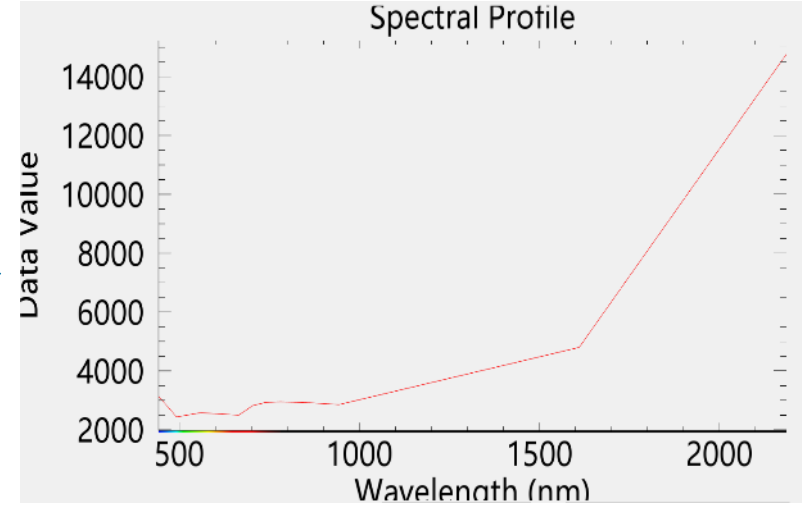
➤ Characteristics Analysis



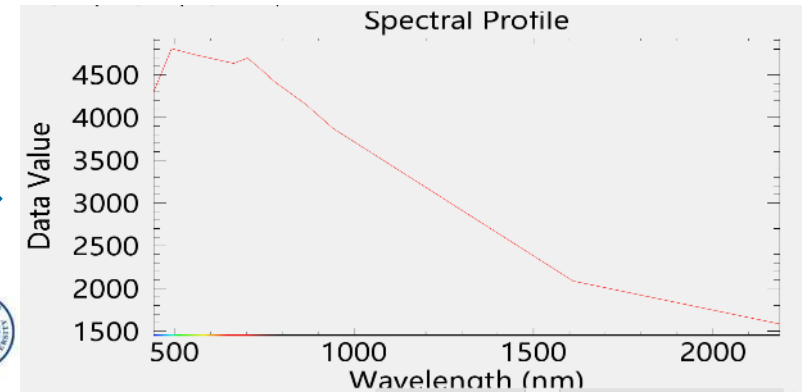
S2B



Fire



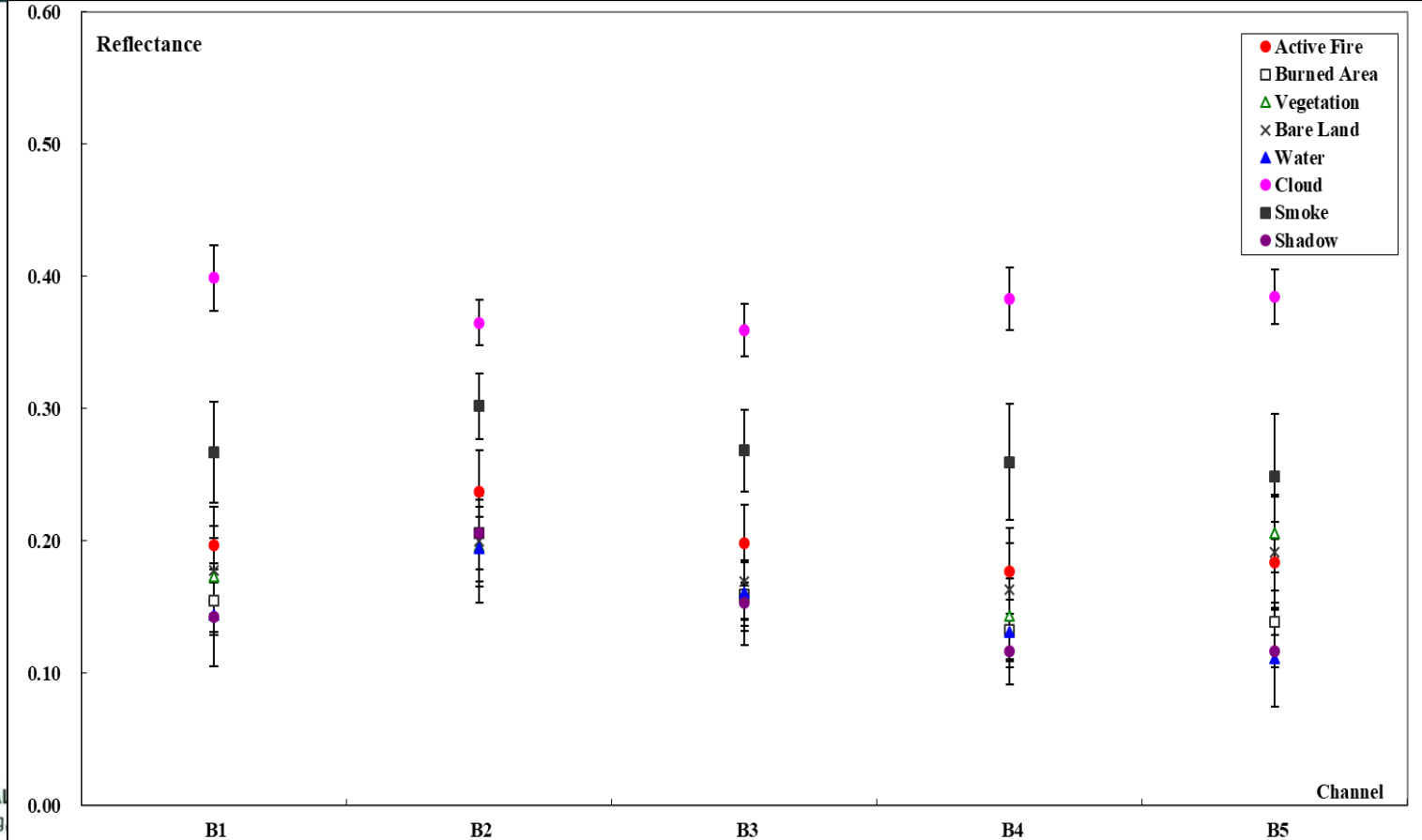
Smoke



Characteristics Analysis



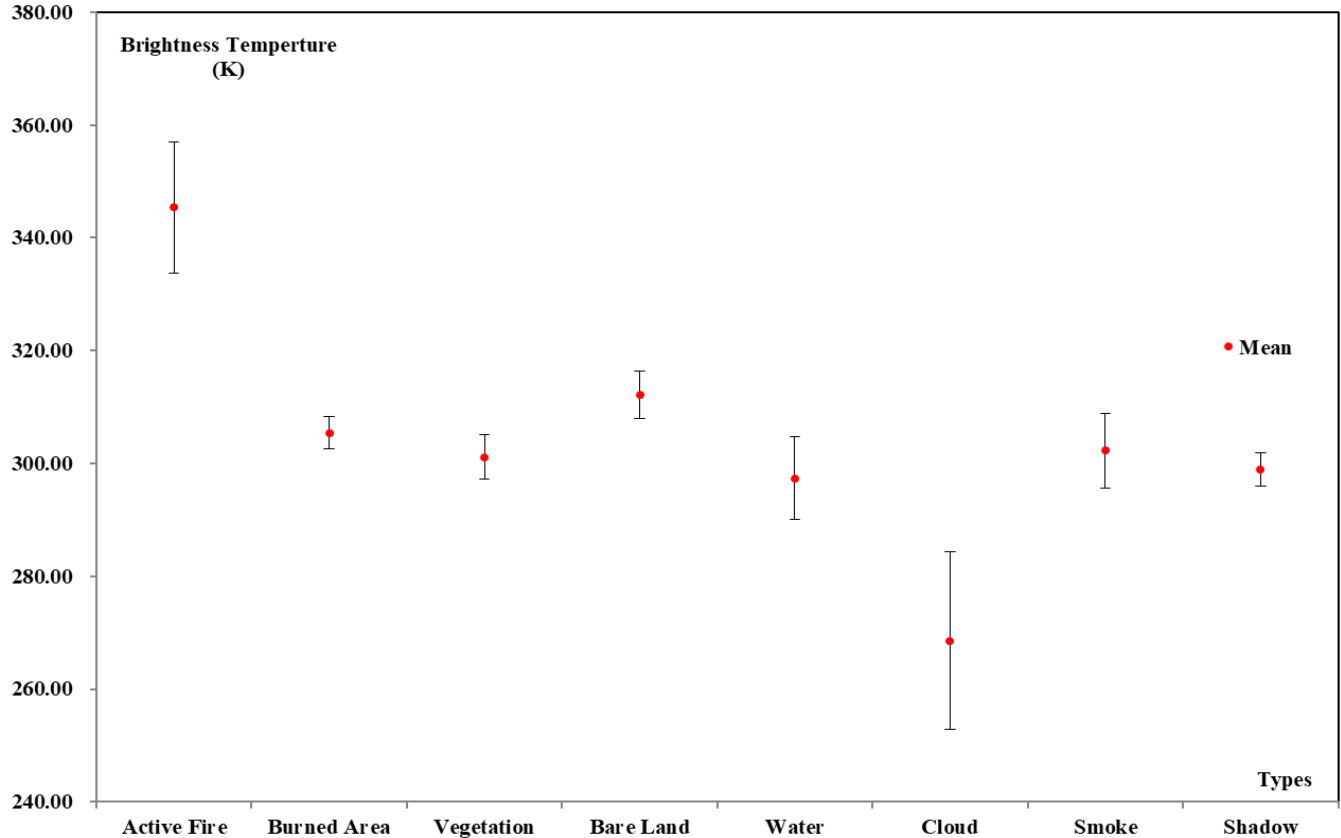
GF-4 PMI



Characteristics Analysis



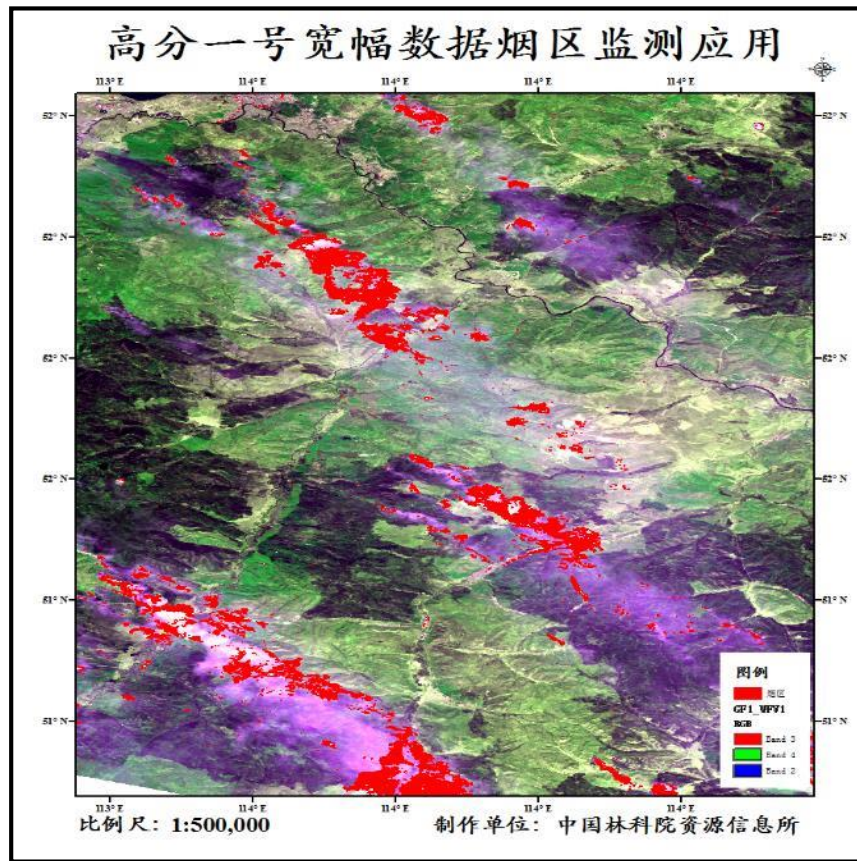
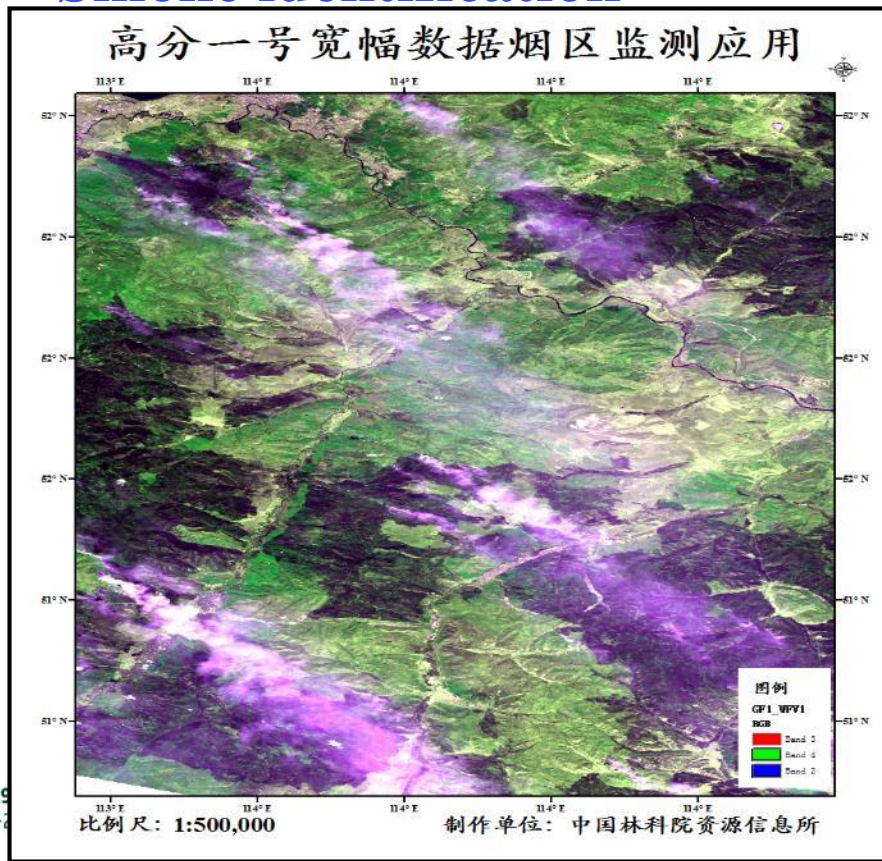
GF-4 PMI



3. Typical Application



➤ Smoke identification



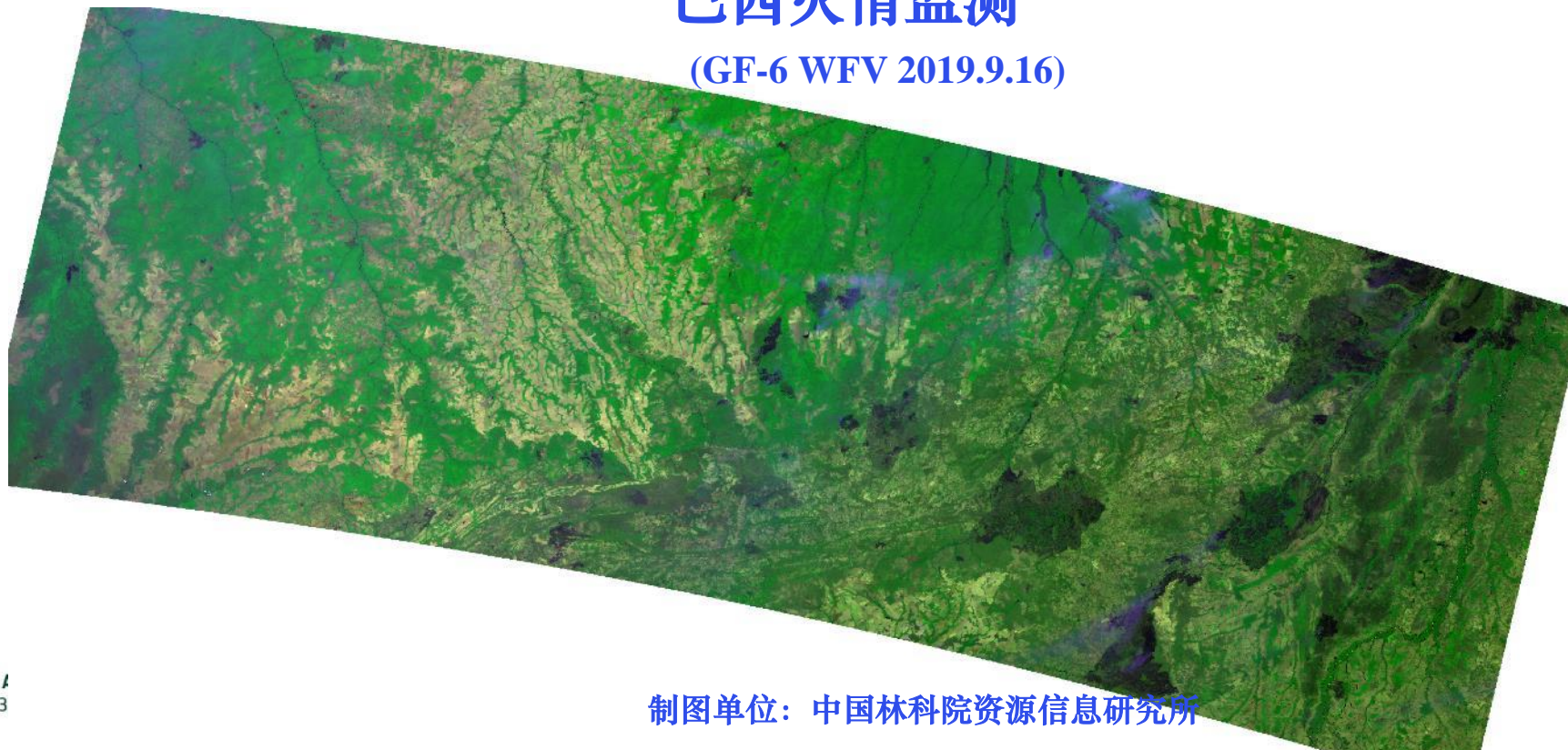
3. Typical Application



➤ Smoke identification

巴西火情监测

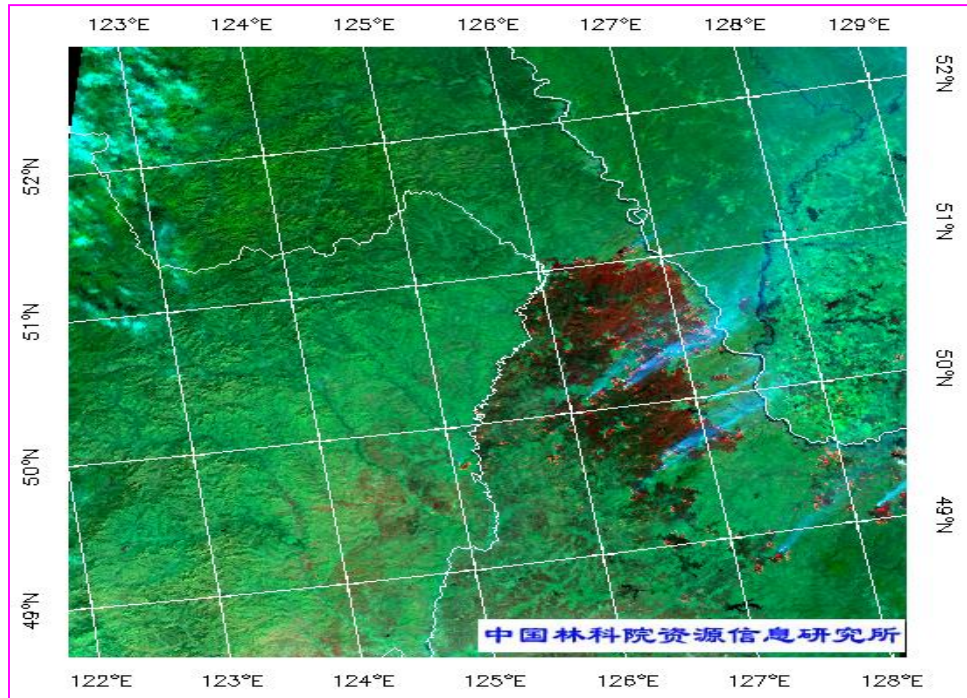
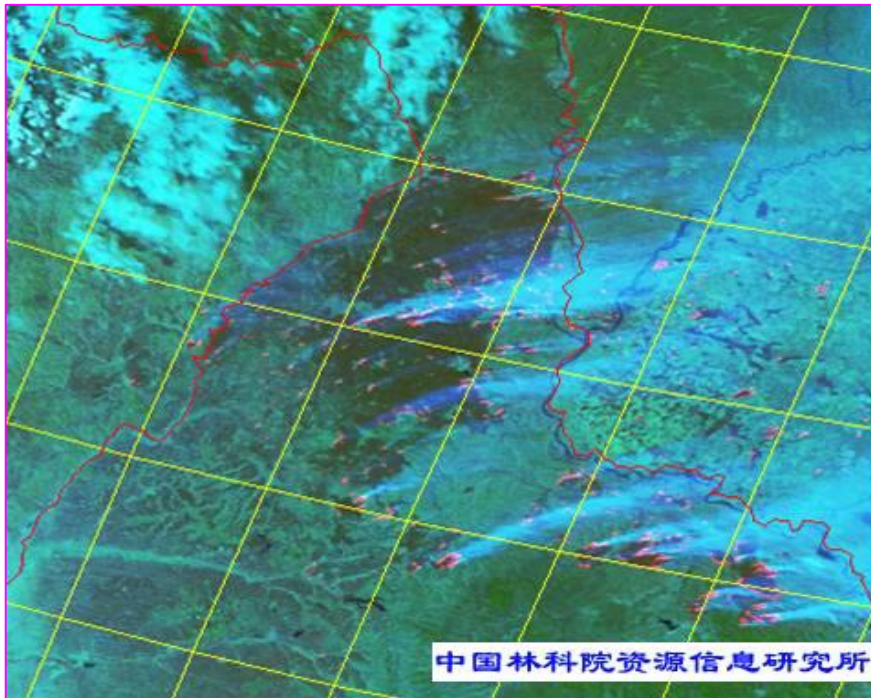
(GF-6 WFV 2019.9.16)



3. Typical Application



➤ Fire identification



NOAA-16

Fire Image, Hei Longjiang Province, 10/17/2004

AATSR

2019 ADVANCED INTERNATIONAL TRAINING COURSE IN LAND REMOTE SENSING
18-23 November 2019 | Chongqing, P.R. China

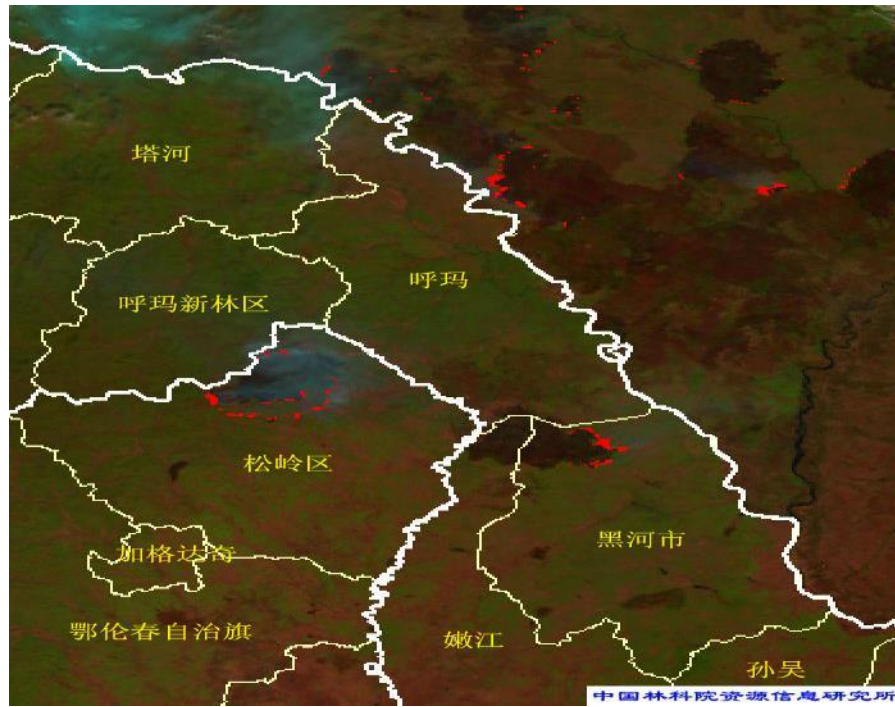
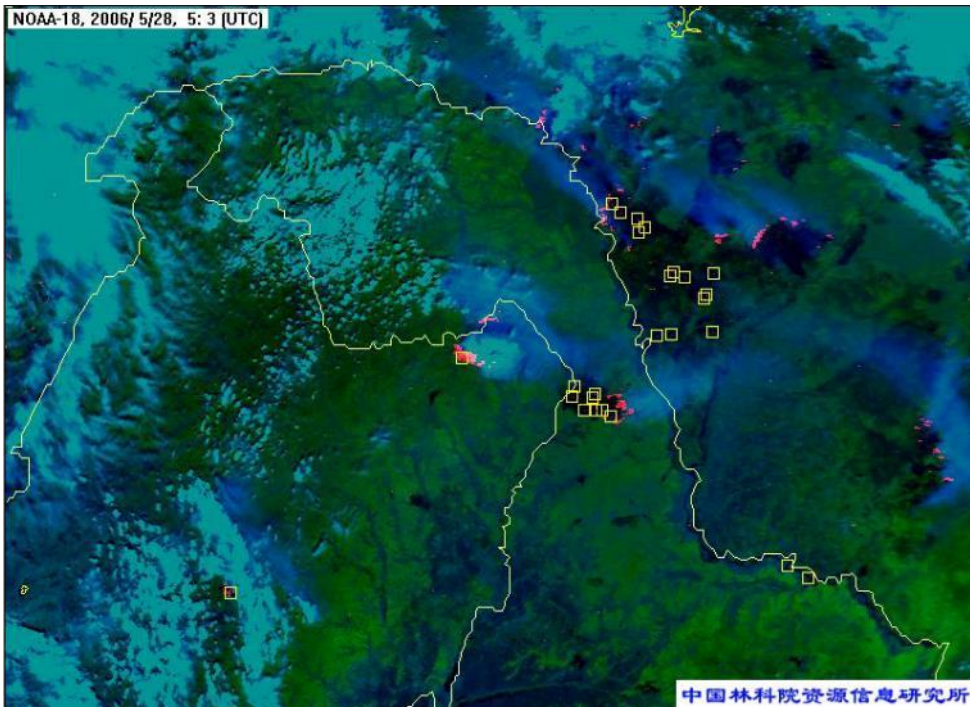


中欧科技合作“龙计划”第四期 2019年陆地遥感高级培训班
培训时间:2019年11月18日-23日 主办方:重庆大学

3. Typical Application



Fire identification



NOAA-18

Fire Image, Inner Mongolia and Hei Longjiang, May 28, 2006

Terra/MODIS

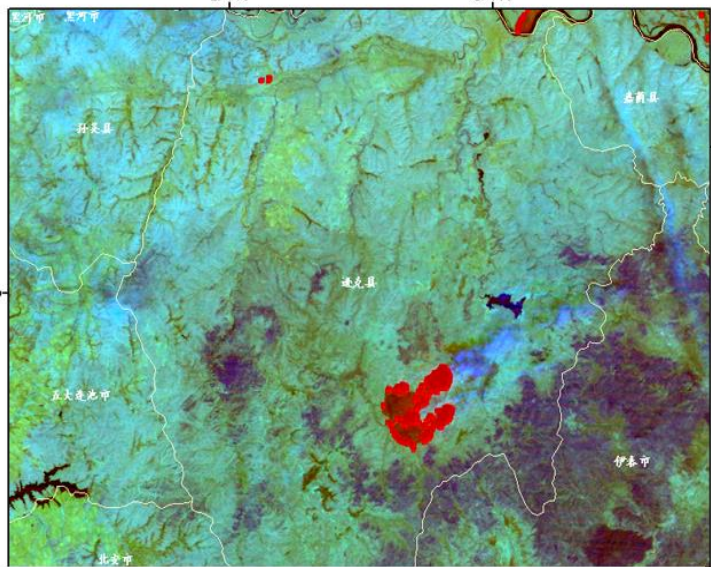


3. Typical Application



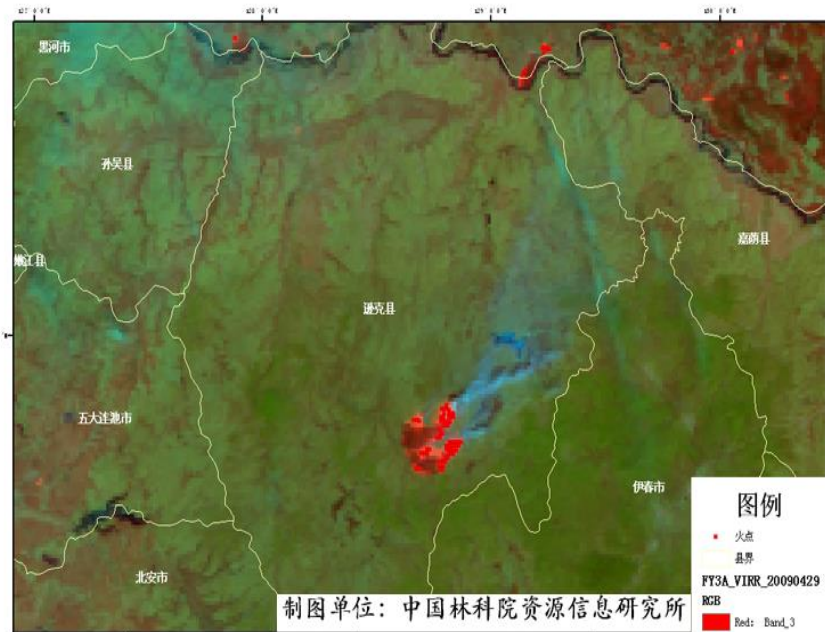
Fire identification

环境减灾小卫星火情监测结果



比例尺: 1:1,000,000 制图单位: 中国林科院资源信息研究所

FY3A VIRR火情监测结果



制图单位: 中国林科院资源信息研究所

Fire Image, Hei Longjiang Province, 4/29/2009



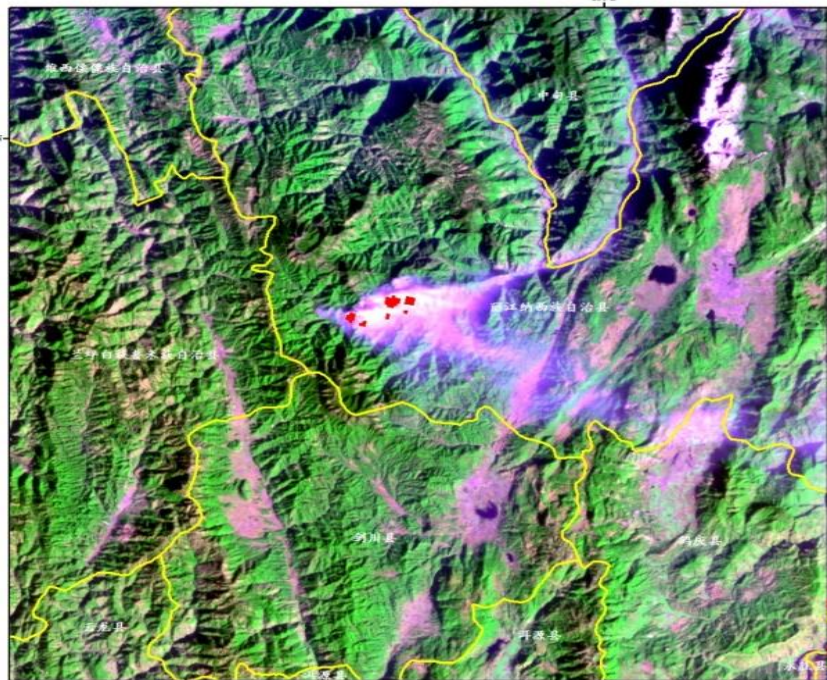
3. Typical Application



Fire identification

高分四号卫星数据火情监测应用

2017年1月21日09:52:40



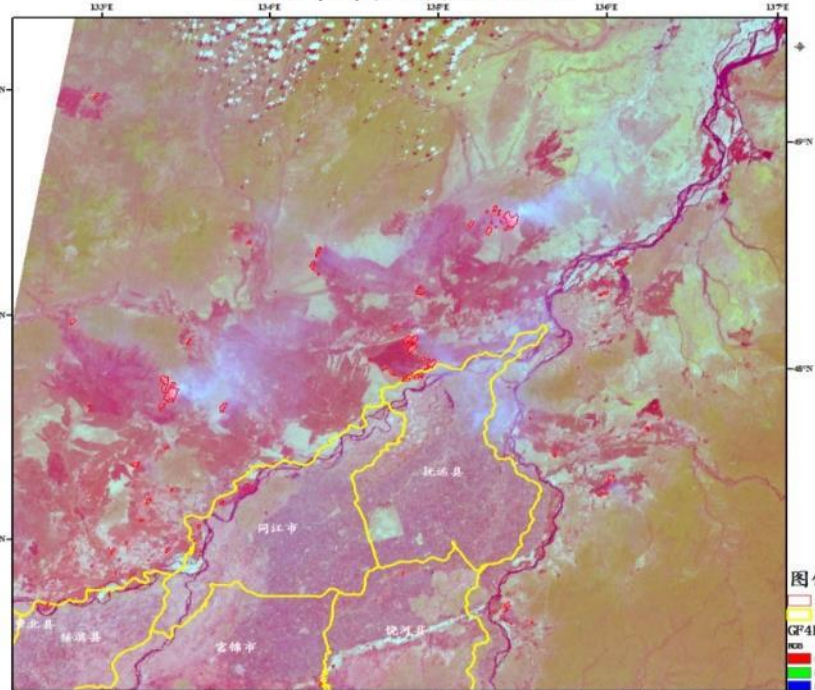
制作单位：中国林科院资源信息研究所

制作时间：2017年1月21日

比例尺：1:500,000

高分四号卫星数据火情监测应用

2018年4月28日11:18:12



制作单位：中国林科院资源信息研究所

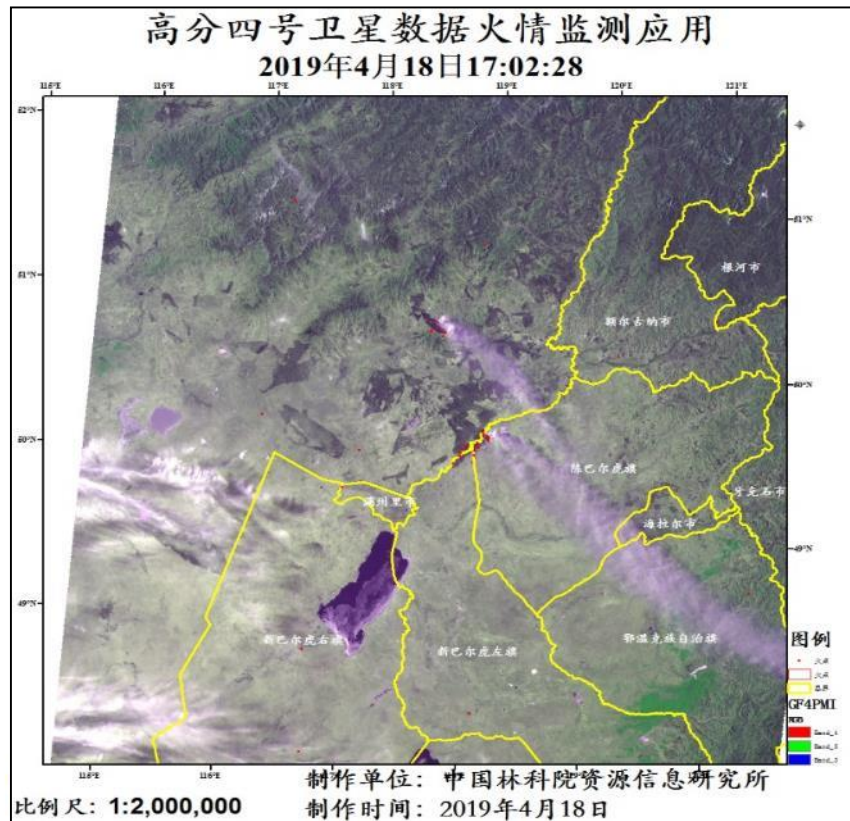
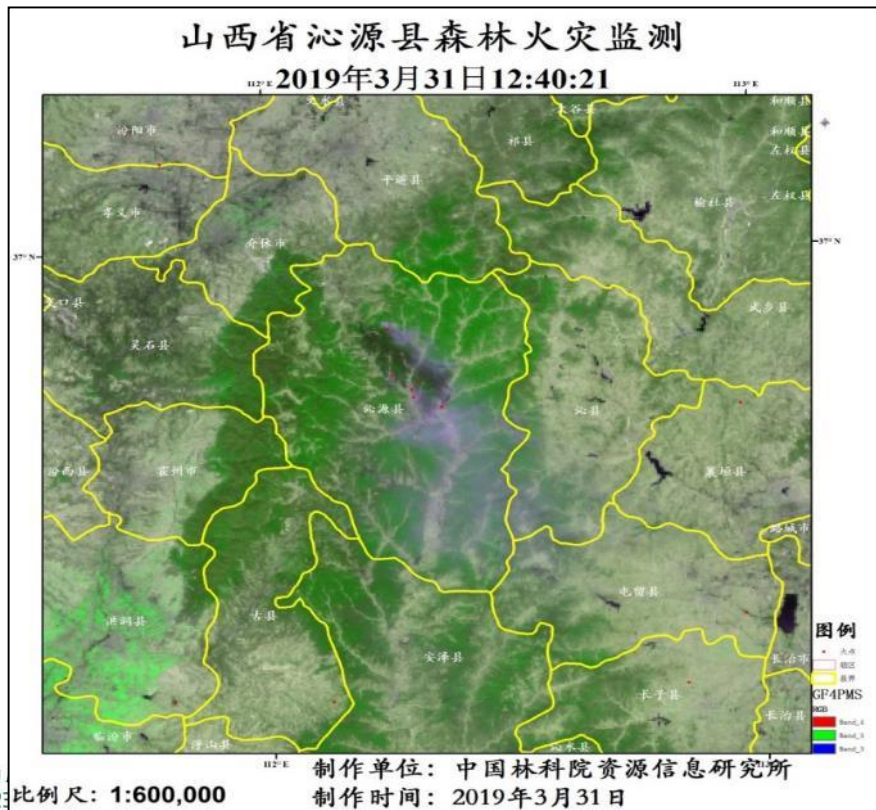
制作时间：2018年4月28日

比例尺：1:1,500,000

3. Typical Application



Fire identification

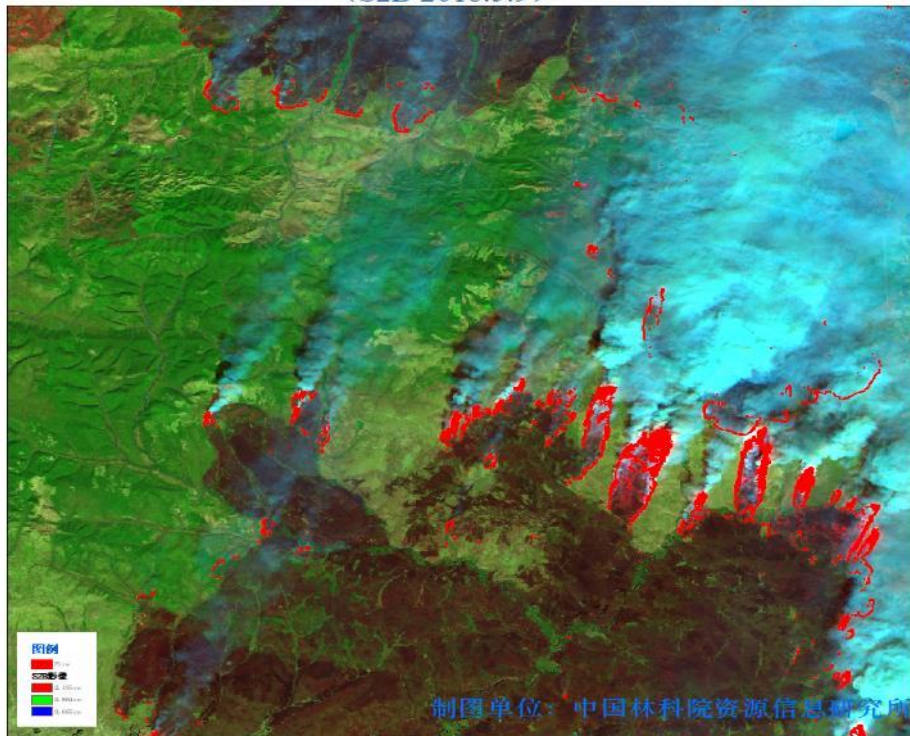


3. Typical Application

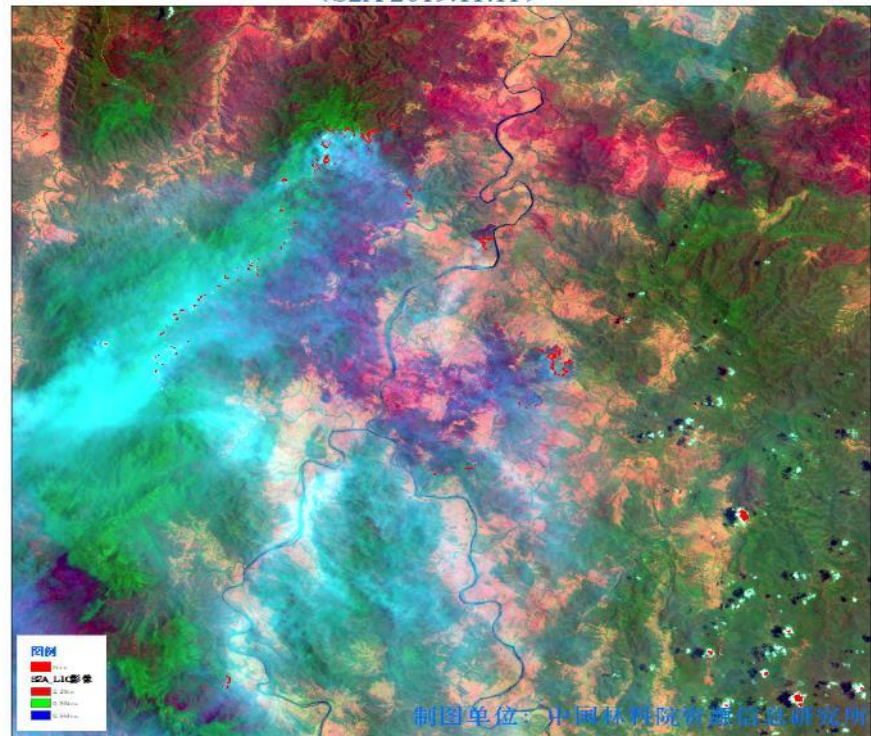


➤ Fire identification

俄罗斯火情监测结果
(S2B 2018.5.9)



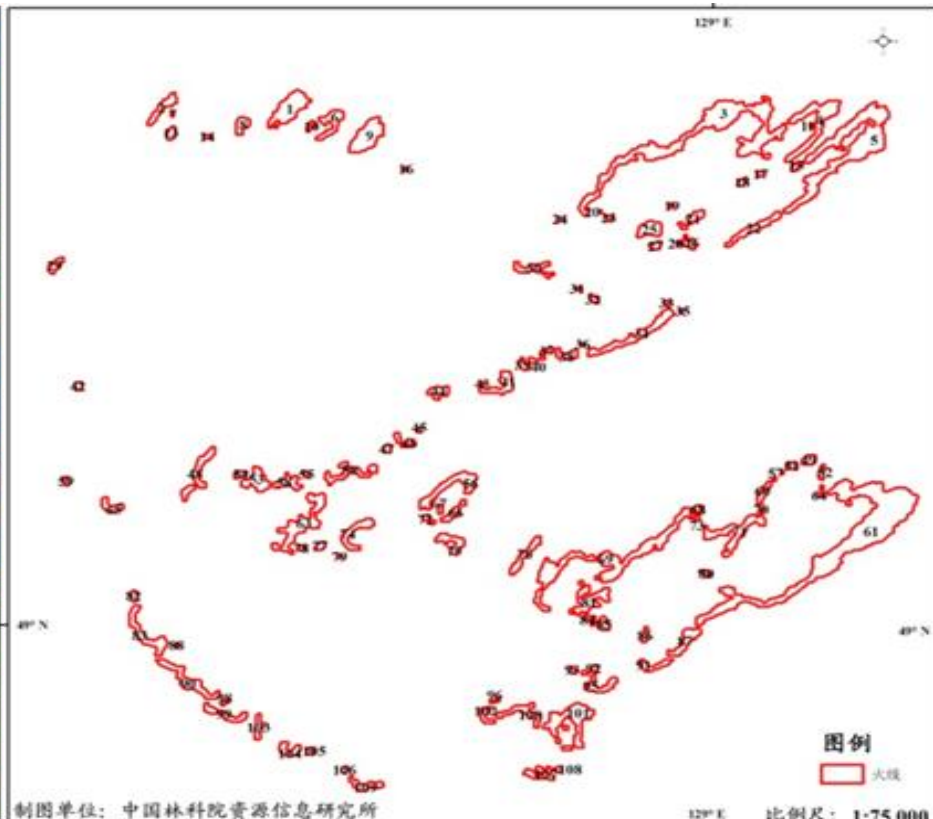
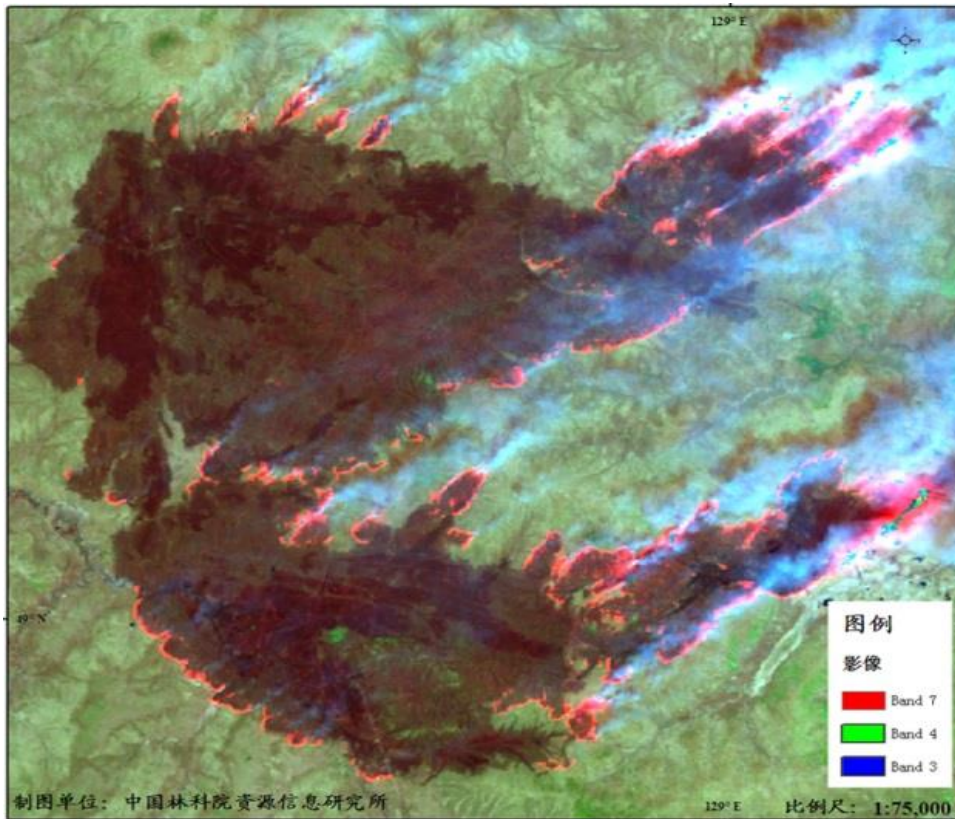
澳大利亚火情监测结果
(S2A 2019.11.11)



3. Typical Application



➤ Fireline extraction

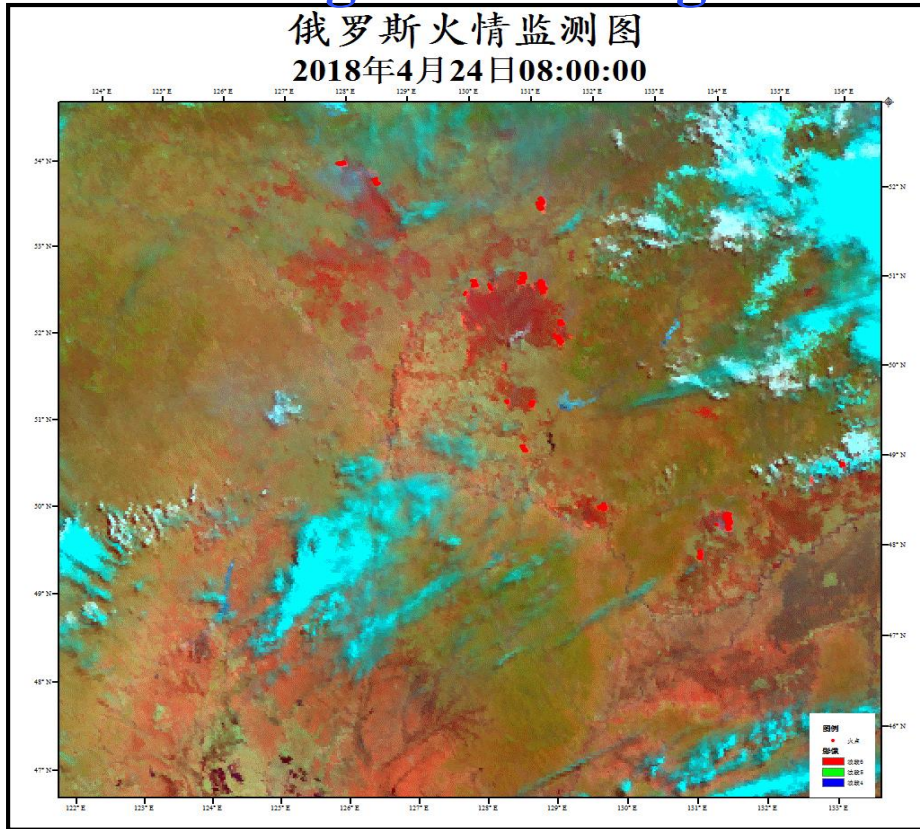


3. Typical Application



➤ Fire change monitoring

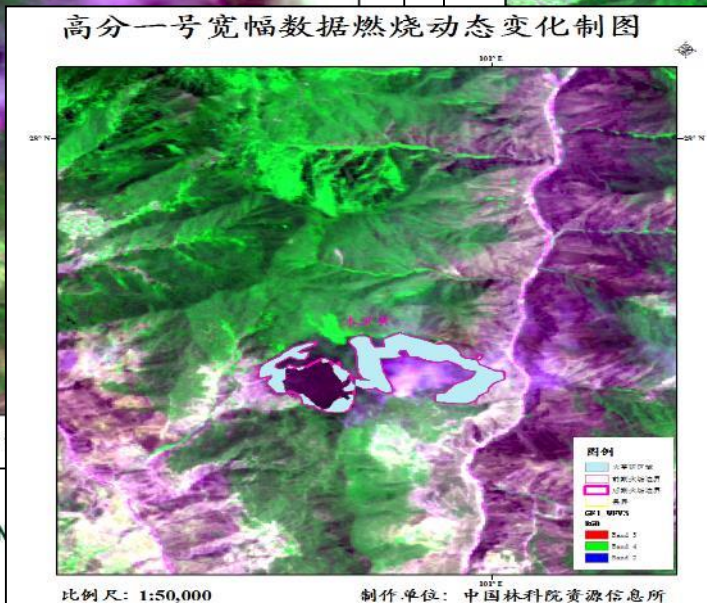
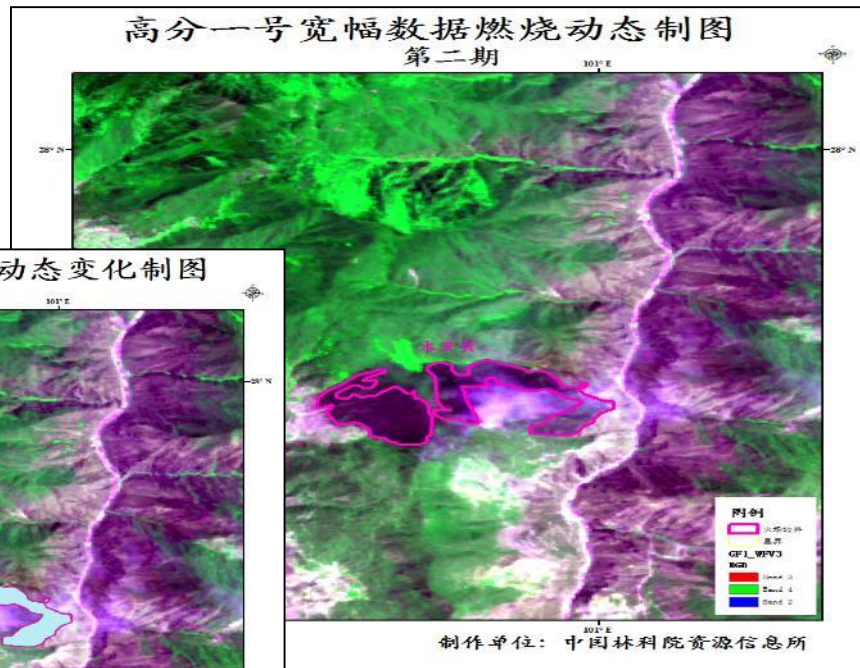
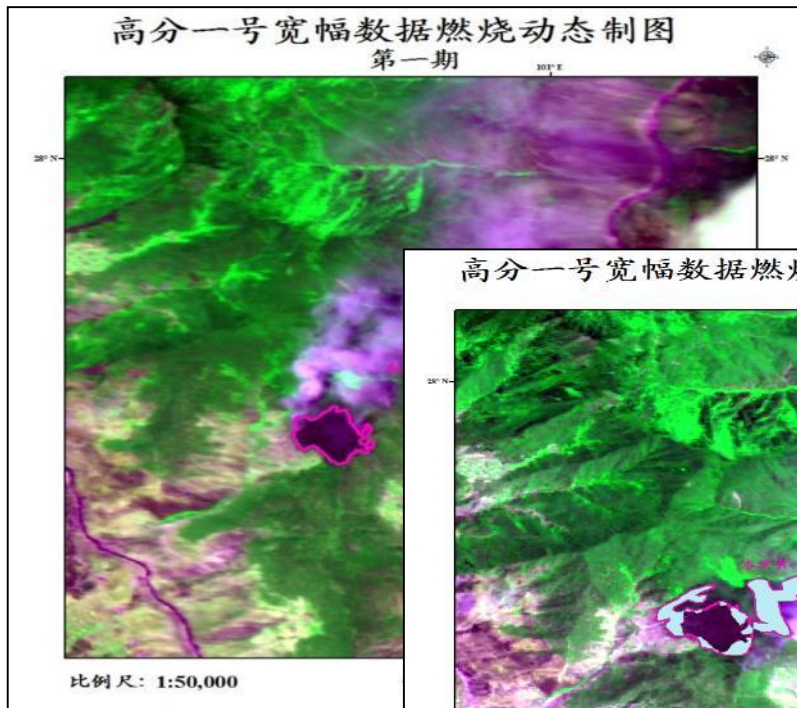
俄罗斯火情监测图
2018年4月24日08:00:00



3. Typical Application



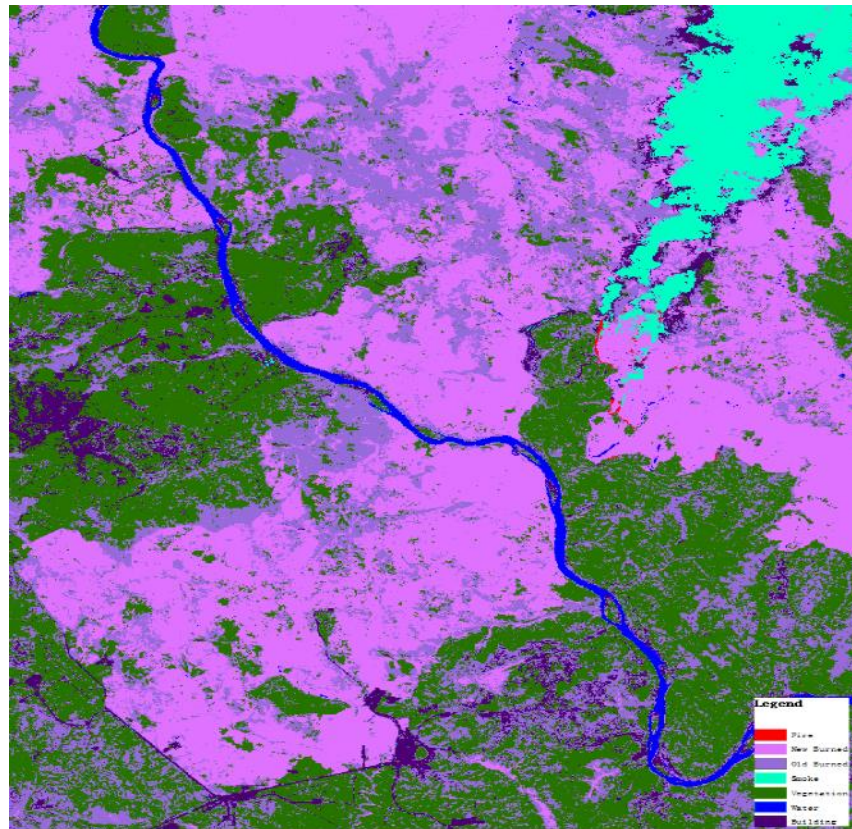
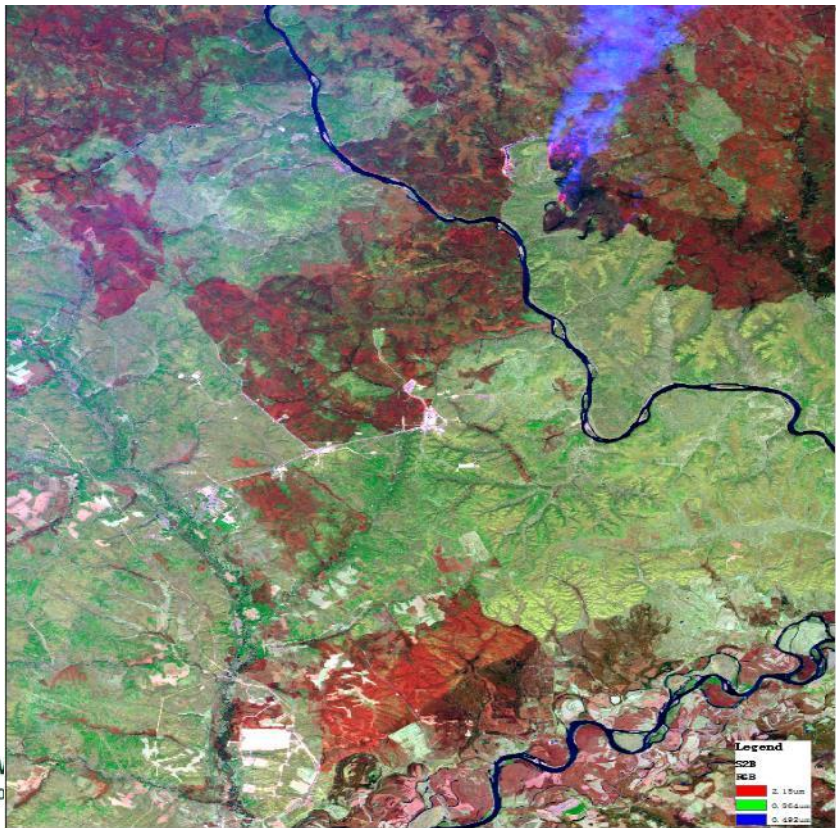
➤ Fire change monitoring



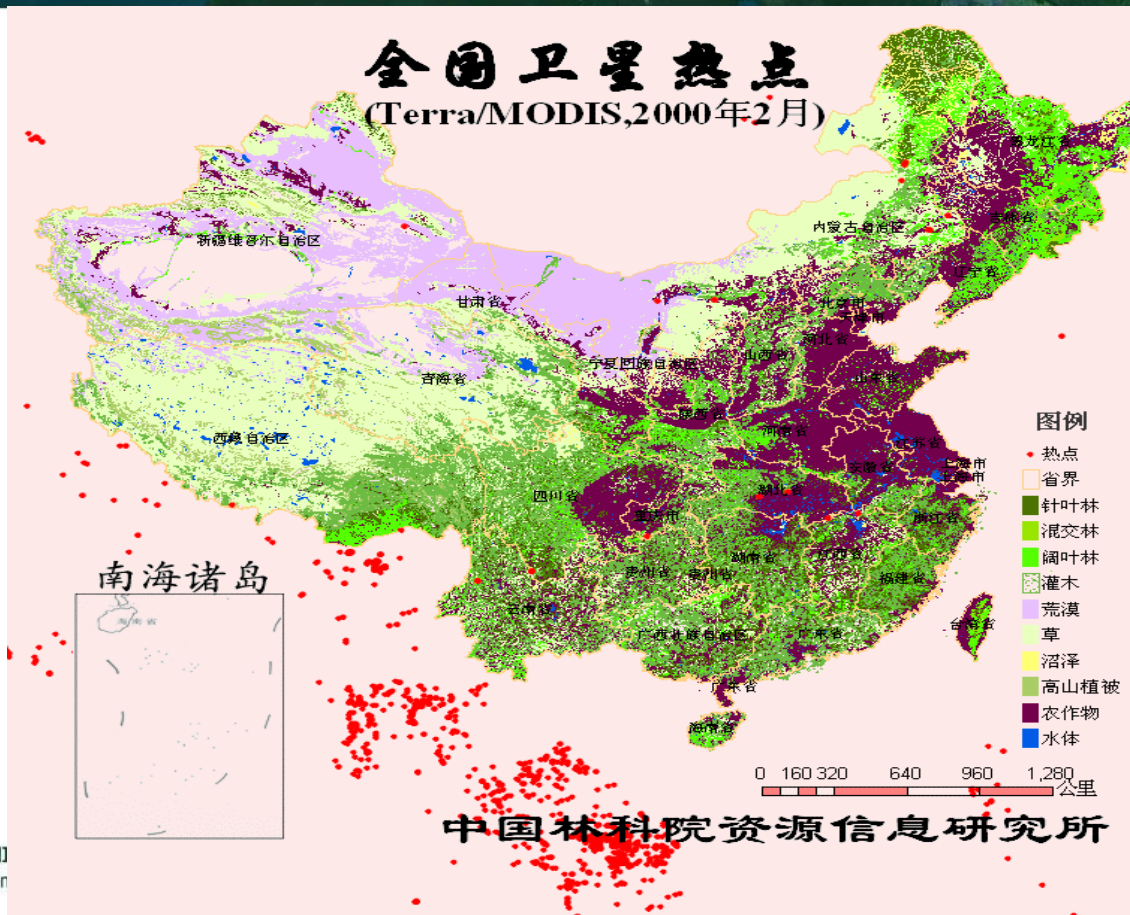
3. Typical Application



➤ Fire status monitoring



3. Typical Application



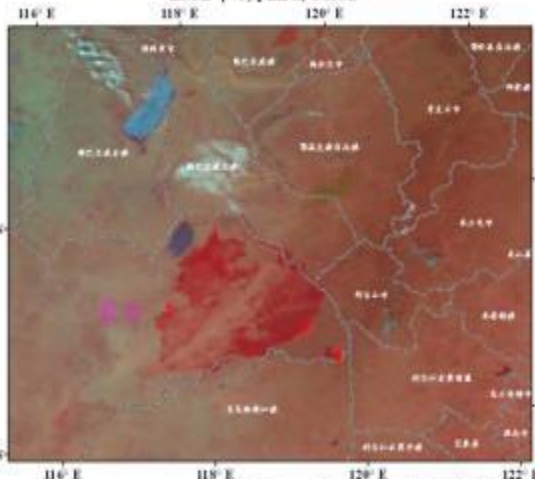
3. Typical Application



科学技术部国家遥感中心

森林火情监测结果

2012年4月22日 10:05



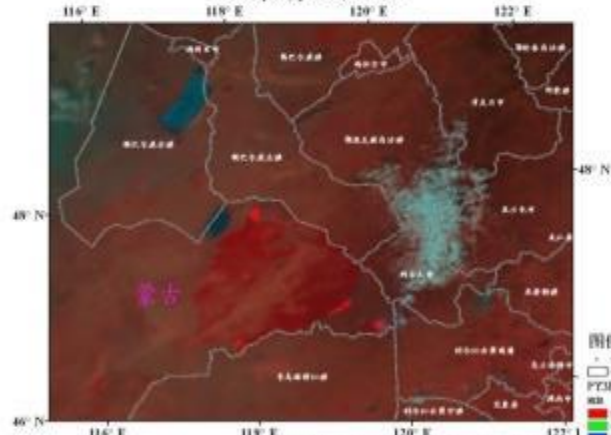
比例尺: 1:2,500,000

制图单位: 中国林科院资源信息研究所
上报时间: 2012年4月22日

科学技术部国家遥感中心

森林火情监测结果

2012年4月22日 12:05



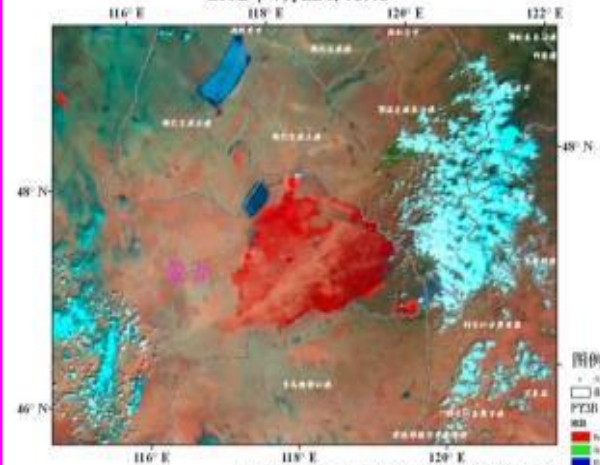
比例尺: 1:2,500,000

制图单位: 中国林科院资源信息研究所
上报时间: 2012年4月22日

科学技术部国家遥感中心

森林火情监测结果

2012年4月22日 13:45



比例尺: 1:2,500,000

制图单位: 中国林科院资源信息研究所
上报时间: 2012年4月22日



Major references



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3. M. J. Wooster, W. Xu, T. Nightingale. Sentinel-3 SLSTR active fire detection and FRP product: Pre-launch algorithm development and performance evaluation using MODIS and ASTER datasets. *Remote Sensing of Environment*, 2012, 120, 236-254.
4. Niels Andela, Douglas C. Morton, Louis Giglio, etc. The global fire atlas of individual fire size, duration, speed and direction. *Earth Syst. Sci. Data*, 2019, 11, 529-552.
5. Wilfrid Schroeder, and Louis Giglio. NASA VIIRS land science investigator processing system (SIPS) Visible Infrared Imaging Radiometer Suite (VIIRS) 375 m & 750 m active fire products, Product User's Guide Version 1.4. July, 2018.

<https://scihub.copernicus.eu/> (Sentinel-1/2/3/5p)





Thanks !

