

# Terra Level 1B Data Fusion Product

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## What is Terra Data Fusion?

### Overview

#### Goals

- Develop a system to efficiently generate mission-scale Terra data fusion products.
- Facilitate the use of Terra data fusion products.

#### Achievements

- Merging L1B radiance fields from all Terra five instruments
- Unpacking radiation data to physical units; and providing geolocation information at native radiance resolutions
- Generating metadata for each Terra orbit
- 2.4 petabytes fusion data generated by supercomputers at NCSA
- HDF5 format, netCDF4-CF compatible
- Tool to resample/reproject fusion data

### Problems

- Large input datasets (~1 PB)
- Input data residing at different locations, in different file formats (HDF 4 & 5), with different granularities
- Inadequate cyberinfrastructure to tackle whole-mission data fusion and mining problems
- Difficult to transfer and manage input data
- Requiring a range of expertise from data scientists to software engineers
- HPC, Data format / interoperability / etc.
- Difficult to use by larger user communities
- Different methods to store radiance and geo-location data for each instrument
- Lack of common metadata conventions

These issues greatly limit and discourage scientists to utilize and analyze the whole Terra data set effectively.

### Solutions

- Utilize expertise and computing infrastructure from different institutions to efficiently generate and deliver Terra data fusion products.
- Fuse radiance data from all instruments.
- Map granules from different instruments to a common granule that contains data for a single Terra orbit.
- Retrieve radiance data that correspond to physical units; data can be ready to use.
- Basic Fusion granules are HDF5 files created from HDF4, HDF-EOS2, and HDF-EOS5 input files. They are netCDF-4 compatible and follow the CF conventions.

## Basic Fusion (BF) – MODIS, MOPITT, CERES, ASTER, MISR

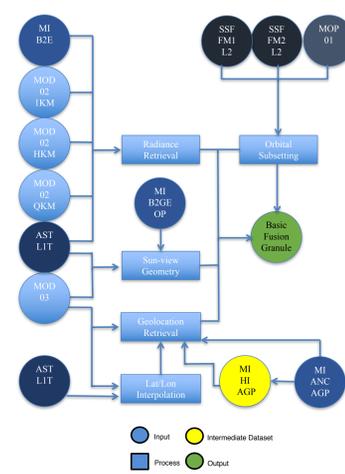
### BF Granularity

- For the Basic Fusion product, a **granularity of one Terra orbit** was chosen.
- A Terra orbit represents a regular and repeating path that the satellite takes around the Earth and is nearly constant over the Terra mission.
- One granule ranges 10GB-70GB, ideal both for IO performance, processing speed, memory usage, and massive data transfer across both disk and tape systems.

### BF File Organization

- The input granules are included if the start time is within the start/end time of the Terra orbit with two exceptions: MOPITT and CERES data are subsetted and then included if any radiance data falls within this orbit.
- Both radiance and geolocation data are included along with sun-view geometry and quality information.
- The BF files were generated in a format that follows CF conventions, such that they can be accessed by netCDF-4 enhanced data models.

### BF Data Generation



### Included Products in NASA CMR\* Short Name and Version Notation

Product	Short Name and Version Notation
MOPITT	MOP01_007
CERES	CER_SSF_Terra-FM1-MODIS_Edition4A_4A CER_SSF_Terra-FM2-MODIS_Edition4A_4A
ASTER	AST_L1T_003
MISR	M1B2E_003, MIB2GEOP_002, MIANCAGP_1, MIHIAGP_1**
MODIS	MOD021KM_6, MOD02HKM_6, MOD02QKM_6, MOD03_6

\*CMR: Common Metadata Repository  
 \*\*275m resolution MISR geolocation fields produced by the Terra fusion project team; unavailable at NASA data centers.

- HDF5 chunking and compression are used to reduce the total BF file size.
- For MODIS, ASTER, and MISR, data converted from native integer (packed) to floating point (unpacked). Floating point data correspond to physical units.
- For MODIS, MISR, and ASTER, geolocation data are interpolated to native radiance resolution and merged into a BF file.
- MOPITT and CERES data are subsetted to span a Terra orbit.

### BF Data Transfer

Instrument	Rate (MB/s)	Size (TB)	DAAC
MISR	160-210	240	ASDC
MODIS	0-100	331	LAADS
ASTER	0-20	400	LPDAAC
CERES	160-210	1.5	ASDC
MOPITT	160-210	0.8	ASDC

- Worked with NASA networking (CNOG), ASDC and LP DAAC teams to diagnose and remedy performance obstacles for data transfer out of the ASDC.
- Result a 2.5x increase in data transfer rates out of the ASDC

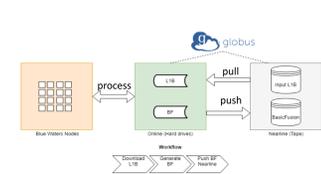
### Lessons Learned

- Mission scale data transfers require active support from the DAACs.
- GridFTP provides much more efficient and reliable protocol than ftp and http for bulk data transfer
- Managing 3+ petabytes of data requires systems specifically designed for big data processing.

### BF Storage and Compute Infrastructure

- Both input and output data are staged on NCSA nearline tape archive system of the BlueWaters
- Input data moved to scratch storage via Globus when generating Basic Fusion files
- NCSA Blue Waters supercomputer used to generate BF data
- TORQUE job and resource management used

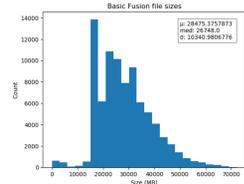
### BF Mission Processing Workflow



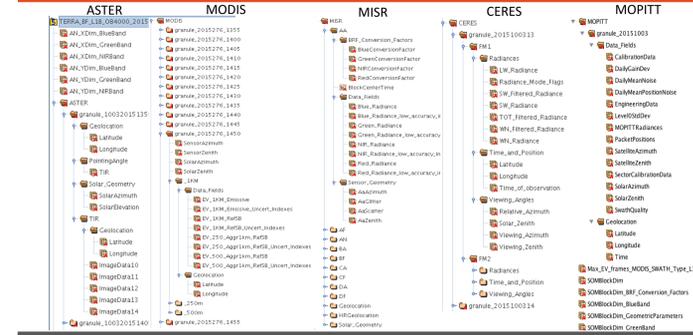
- 130,000 core CPU hours are used for actual mission processing.
- One granule takes 1-2 hours to produce.
- C is used for Basic Fusion software and Python for workflow.
- Granules are produced in batches of 2,000 to 3,000.
- Smaller NCSA cluster (ROGER) was used during code development and testing.
- SQLite database was used to query input granules by orbit.

### BF File Statistics

- 84,303 files - from 02/25/2000 to 12/31/2015
- The total file size is 2.4 petabytes.
- Typical file sizes 15GB – 40GB. The largest file size is 68.7GB. Average file size is 26GB.
- In-memory compression reduces the total file size by 60%. Without using compression, the total size will be about 5.5–6 petabytes.
- The biggest BF file takes about 3 hours to finish at NCSA Blue Waters.



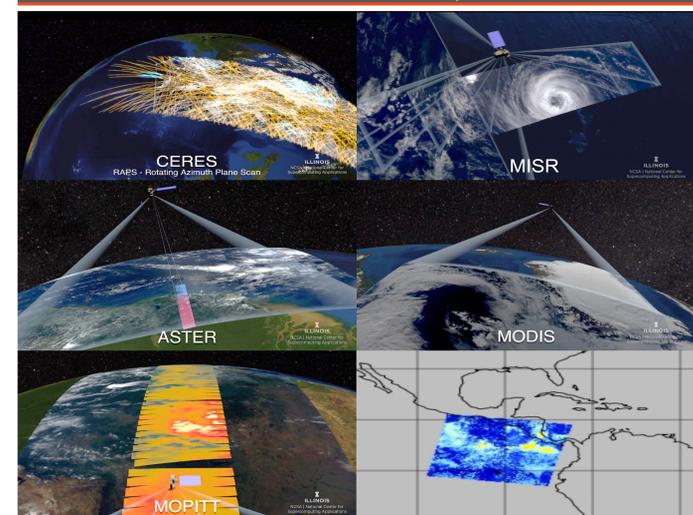
### BF HDF5 File Layout in HDFView



### BF HDF5 File Layout in CDL

```
group: ASTER {
  group: TIR {
    variables:
    float ImageData10(ImageLine_TIR_Swath_g10222015151007,
    ImagePixel_TIR_Swath_g10222015151007);
    ImageData10.FillValue = 9999.f;
    ImageData10.valid_min = 0.f;
    ImageData10.valid_max = 569.f;
    ImageData10.coordinates = "ASTER/granule_10222015151007/TIR/Geolocation/Longitude/ASTER/granule_10222015151007/TIR/Geolocation/Latitude";
  }
  group: MISR {
    variables:
    float Blue_Radiance(SOMBlockDim_BlueBand_XDim_BlueBand_YDim_BlueBand);
    Blue_Radiance.coordinates = "MISR/Geolocation/GeoLongitude/MISR/Geolocation/GeoLatitude";
    Blue_Radiance.units = "MISR/2micrometersteradian";
    Blue_Radiance.FillValue = 999.f;
    Blue_Radiance.valid_min = 0.f;
    Blue_Radiance.valid_max = 800.f;
  }
}
group: CERES {
  group: granule_2015100214 {
    group: FM1 {
      group: Radiance {
        variables:
        float LW_Radiance(Footprints_FM1_g2015100214);
        LW_Radiance.FillValue = 3.402823e+38f;
        LW_Radiance.valid_range = 0.f, 200.f;
        LW_Radiance.units = "Watts per square meter per steradian";
      }
      group: Time_and_Position {
        variables:
        float Latitude(Footprints_FM1_g2015100214);
        Latitude.units = "degrees_north";
        float Longitude(Footprints_FM1_g2015100214);
        Longitude.units = "degrees_east";
        double Time_of_observation(Footprints_FM1_g2015100214);
        Time_of_observation.units = "day";
      }
    }
  }
}
group: MOPITT {
  group: granule_20151002 {
    group: Data_Fields {
      variables:
      float MOPITTRadiance(track_1_nstare, npixels, nchan, nstare);
      MOPITTRadiance.coordinates = "MOPITT/granule_20151002/Geolocation/Latitude/MOPITT/granule_20151002/Geolocation/Longitude";
      MOPITTRadiance.FillValue = 9999.f;
      string MOPITTRadiance.units = "Watts/m^2steradian";
    }
    group: Geolocation {
      variables:
      float Latitude(track_1_nstare, npixels);
      Latitude.units = "degrees_north";
      float Longitude(track_1_nstare, npixels);
      Longitude.units = "degrees_east";
      double Time(track_1);
      string Time.units = "seconds since 1993-01-01";
    }
  }
}
group: MODIS {
  group: 1KM {
    group: Data_Fields {
      variables:
      float EV_1KM_Emissive(Band_1KM_Emissive_MODIS_SWATH_Type_L1B,
      _10_nstars, MODIS_SWATH_Type_L1B, Max_EV_frames_MODIS_SWATH_Type_L1B);
      EV_1KM_Emissive.valid_min = 0.f;
      EV_1KM_Emissive.units = "Watt/m^2micrometersteradian";
      EV_1KM_Emissive.coordinates = "MODIS/granule_2015295_1425_1KM/Geolocation/Longitude/MODIS/granule_2015295_1425_1KM/Geolocation/Latitude";
      string EV_1KM_Emissive.band_names = "20.21.22.23.24.25.27.28.29.30.31.32.33.34.35.36";
      float EV_1KM_Emissive_Uncorr_Index(Band_1KM_Emissive_MODIS_SWATH_Type_L1B,
      _10_nstars, MODIS_SWATH_Type_L1B, I_Max_EV_frames_MODIS_SWATH_Type_L1B);
    }
  }
}
// group Data_Fields
}
```

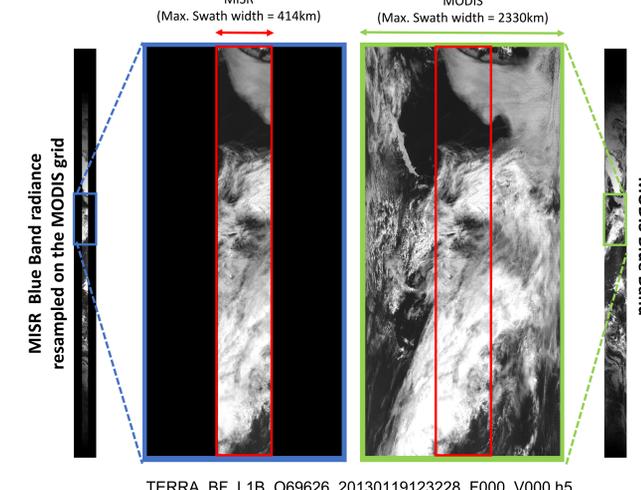
### BF Visualization Examples



## Advanced Fusion (AF)

### AF Resampling and Reprojection Tool (in development)

- A community, open source tool
  - Resample / reproject the radiance fields for one Terra instrument onto the grid used by another Terra instrument, a regularly spaced grid, or any other user specified grid.
- Two major resampled strategies implemented (Nearest neighbor resampling & Statistical resampling)
- User-specified map projections and features
  - Regularly spaced grids defined by GDAL-supported map projections
  - MISR block unstacking
  - Common swath selection
- HDF5 and GeoTIFF output
- 40 seconds to resample MISR 1.1km radiance data to MODIS 1.0 km grid for an entire orbit using 32 threads on Blue Waters



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## Other Related Work

### Metadata Generation

- Collection and granule metadata in ECHO10 XML are available.
- Although input data are missing for some orbits, granule metadata are generated to help users locate these orbits through NASA CMR search.
- The metadata standards and format passed the CMR ingestion test

### OPeNDAP Web Service

- BF files are tested and can be accessed with the OPeNDAP Hyrax server.
- OPeNDAP subset service may help users to retrieve data easily and efficiently.

### Additional Tools

- A Globus Python interface data transfer tool
- A Python PBS workflow manager

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