

NOAA/NSIDC Climate Data Record of Passive Microwave Sea Ice Concentration, Version 4

USER GUIDE

How to Cite These Data

As a condition of using these data, you must include a citation:

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FOR QUESTIONS ABOUT THESE DATA, CONTAC[T NSIDC@NSIDC.ORG](mailto:nsidc@nsidc.org)

FOR CURRENT INFORMATION, VISI[T https://nsidc.org/data/G02202/](https://nsidc.org/data/G02202/)

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1 DATA DESCRIPTION

Notice: A near-real-time version of this data set also exists to fill the gap between the time that this data set is updated through to the present. The data set is called the *[Near-Real-Time NOAA/NSIDC Climate](https://nsidc.org/data/g10016) [Data Record of Passive Microwave Sea Ice Concentration](https://nsidc.org/data/g10016)* (G10016).

1.1 Summary

This data set provides passive-microwave-derived sea ice concentration (SIC) estimates that are produced in conformance with NOAA Climate Data Record (CDR) program criteria (NRC 2004). These criteria emphasize transparent and reproducible processing. The SIC CDR algorithm output is a rule-based combination of ice concentration estimates from two well-established SIC algorithms: the NASA Team (NT) algorithm (Cavalieri et al. 1984) and NASA Bootstrap (BT) algorithm (Comiso 1986). The SIC CDR algorithm blends the NT and BT output concentrations by selecting, for each grid cell, the higher concentration value. It capitalizes on the strengths of each contributing algorithm to produce ice concentration fields that should be more accurate than those from either algorithm alone. This statement is based on SIC CDR algorithm logic and the literature of NT and BT validation studies. Comprehensive validation of CDR ice concentration fields has not taken place. However, Meier et al. (2014) provide a detailed analysis of the spatial distributions of differences between the SIC CDR fields and ice concentration from NT and BT. They find that the CDR and BT fields are quite similar in both hemispheres. There are larger differences between CDR and NT, with the CDR (and BT) finding more ice overall. Trends in area and extent for all three products, computed over 1988-2007, have only small differences. This document summarizes important information about this data set including data file information and organization, spatial and temporal resolution, and data acquisition and processing. For full details on the algorithms, filters, interpolations, and error sources, see the Climate Algorithm Theoretical Basis Document (C-ATBD): Sea Ice Concentration (Meier et al., 2021).

The NT and BT algorithms run at NSIDC as part of SIC CDR processing. Separately, NASA Goddard Space Flight Center (GSFC) produces ice concentrations using the NT and BT algorithms that are distributed by the NSIDC DAAC as the following data sets: *[Sea Ice Concentrations from](https://nsidc.org/data/nsidc-0051) [Nimbus-7 SMMR and DMSP SSM/I-SSMIS Passive Microwave Data](https://nsidc.org/data/nsidc-0051)* (NSIDC-0051) and *[Bootstrap](https://nsidc.org/data/nsidc-0079) [Sea Ice Concentrations from Nimbus-7 SMMR and DMSP SSM/I-SSMIS](https://nsidc.org/data/nsidc-0079)* (NSIDC-0079), respectively. These products have some manual quality control applied and therefore do not meet CDR standards for reproducibility.

The SIC CDR begins in 1978 with NASA Nimbus-7 SMMR data with the variables cdr_seaice_conc for daily and cdr_seaice_conc_monthly for monthly. This data set is updated approximately every three to six months, after the input brightness temperature data (NSIDC-0001) become available. The closely related *[Near-Real-Time NOAA/NSIDC Climate Data Record of](https://nsidc.org/data/g10016) [Passive Microwave Sea Ice Concentration](https://nsidc.org/data/g10016)* fills the gap between the end date of this data set and present.

Daily and monthly resolution sea ice concentration values are provided in NetCDF files organized in two ways: 1) one file for each day of the year and one file for each month of the year for each hemisphere and 2) daily data aggregated into yearly files and monthly data aggregated into one period-of-record file for each hemisphere. Each file has a variable for the concentration product, as well as variables containing standard deviation, quality flags, and projection information. All data are on a 25 km x 25 km grid.

1.2 Parameters

The parameter of this data set is sea ice concentration which is the fraction of ocean area covered by sea ice. Sea ice concentration represents an areal coverage of sea ice. For a given grid cell, the parameter provides an estimate of the fractional amount of sea ice covering that cell, with the remainder of the area consisting of open ocean. Land areas are coded with a land mask value.

1.3 File Information

1.3.1 Format

These data are provided in NetCDF4 file format and are compliant with the Climate and Forecast (CF) Metadata Convention CF-1.6 (Eaton et al., 2010) and the Attribute Convention for Data Discovery (ACDD) 1.3.

The variables in both the daily and monthly NetCDF files are described in the sections [1.3.2.1](#page-3-4) [Daily](#page-3-4) [File Variable Description](#page-3-4) and [1.3.2.2](#page-12-0) [Monthly File Variable Description,](#page-12-0) respectively.

1.3.2 File Contents

1.3.2.1 Daily File Variable Description

The daily NetCDF4 files contain the variables listed in [Table 1,](#page-4-0) which provides a brief description of each. The sections below this table provide more detailed information.

Table 1. Daily Variables at a Glance. Click Variable Name for More Information.

cdr_seaice_conc

latitude

longitude

melt_onset_day_cdr_seaice_conc

Description Contains the day of year on which melting sea ice was first detected in each grid cell. Once detected, the value is retained for the rest of the year. For example, if a grid cell started melting on day 73, the variable for the grid cell on that day will be 73, as will all subsequent days until the end of the year. The melt onset day is only calculated for the melt season: days 60 through 244, inclusive. Before melting is detected or if melt is never detected for that grid

nsidc_bt_seaice_conc

nsidc_nt_seaice_conc

qa_of_cdr_seaice_conc

Table 3. Daily QA Flag Values

spatial_interpolation_flag

Condition	Flag Value	Label in NetCDF Variable
19 GHz vertical brightness temperature spatially interpolated	1	19v tb value interpolated
19 GHz horizontal brightness temperature spatially interpolated	$\overline{2}$	19h tb value interpolated
22 GHz vertical brightness temperature spatially interpolated	$\overline{4}$	22v tb value interpolated
37 GHz vertical brightness temperature spatially interpolated	8	37v tb value interpolated
37 GHz horizontal brightness temperature spatially interpolated	16	37h tb value interpolated
Pole hole spatially interpolated (Arctic only)	32	pole hole value interpolated

Table 4. Spatial interpolation flag values. A grid cell that satisfies more than one criteria will contain the sum of all applicable flag values.

stdev_of_cdr_seaice_conc

temporal_interpolation_flag

time

xgrid

1.3.2.2 Monthly File Variable Description

The monthly NetCDF4 files contain the variables listed in [Table 5,](#page-12-2) which provides a brief description of each. The sections below this table provide more detailed information.

cdr_seaice_conc_monthly

latitude

longitude

melt_onset_day_cdr_seaice_conc_monthly

nsidc_bt_seaice_conc_monthly

nsidc_nt_seaice_conc_monthly

projection

qa_of_cdr_seaice_conc_monthly

The QA flags listed in Table 6 include whether the average concentration exceeds 15%, which is commonly used to de[fine the ic](#page-16-0)e edge and can be used to easily quantify the total extent. Another flag indicates when average concentration exceeds 30%, which is a commonly used alternate ice edge definition. It may be desired to remove lower concentration ice that tends to have higher errors. Another flag indicates whether at least half the days have a concentration greater than 15%. This provides a monthly median extent, which may be a better representation of the monthly ice presence because an average conflates the spatial and temporal variation through the month. Additionally, there is a flag that indicates whether at least half the days have a concentration greater than 30%. This also provides a monthly median extent, but this higher percentage may leave out questionable or erroneous ice. There are flags to show if a cell was masked by the valid ice mask and whether spatial or temporal interpolation was performed. Finally, there is a flag to note whether melt was detected during the month. Since melt tends to bias concentrations lower, this flag gives a sense of whether melt has any effect on the monthly concentration estimate and whether it is having a dominating effect.

Table 6. Monthly QA Flag Values

stdev_of_cdr_seaice_conc_monthly

time

xgrid

1.3.2.3 Ancillary Files

Two ancillary files accompany this data set, one for the Northern Hemisphere and one for the Southern Hemisphere: G02202-cdr-ancillary-nh.nc and G02202-cdr-ancillary-sh.nc. These files contain the land mask, latitude, longitude, minimum concentration mask, pole hole masks, and valid ice masks used in processing the sea ice CDR. [Table 7](#page-18-1) describes the contents of these files.

Table 7. Ancillary file description

1.3.3 Directory Structure

The data files are organized on the HTTPS site into two main directories by hemisphere: north and south. The top-level directory also contains an ancillary directory that holds ancillary data files that may be useful when working with the sea ice CDR. Within each of the hemisphere directories, there are four sub-directories: aggregate, checksums, daily, and monthly. The aggregate directory contains the yearly aggregated daily files and the period-of-record aggregated monthly files. The checksums directory contains md5 checksums of the individual daily and monthly data files and the aggregated daily and monthly data files to ensure accuracy in data transfer. The daily directory contains the individual daily data files and is further sub-divided into directories labeled by the 4-digit year (YYYY) beginning with 1978; the daily files reside within their respective year directory. All individual monthly files reside directly in the monthly directory.

1.3.4 Naming Convention

The file naming convention for the daily and monthly files is listed below and described in [Table 8:](#page-19-2)

Individual daily files: seaice_conc_daily_hh_yyyymmdd_sat_vXXrXX.nc Yearly aggregated daily files: seaice_conc_daily_hh_yyyy_vXXrXX.nc Individual monthly files: seaice_conc_monthly_hh_yyyymm_sat_vXXrXX.nc Period-of-record aggregated monthly files: seaice_conc_monthly_hh_197811_yyyymm_vXXrXX.nc

Where:

Variable	Description
seaice conc	Identifies files containing sea ice concentration data
daily	Identifies files containing daily sea ice concentration
monthly	Identifies files containing monthly sea ice concentration
hh	Hemisphere (nh: North, sh: South)
yyyy	4-digit year
mm	2-digit month
dd	2-digit day of month
sat	Satellite the data came from (n07: Nimbus 7, f08: DMSP F8, f11: DMSP F11, f13: DMSP F13, f17: DMSP F17)
vXXrXX	Version and revision number of the data file (v04r00: Version 4, Revision 0)
.nc	Identifies a NetCDF file
.nc.mnf	Identifies this as an md5 checksum file

Table 8. File Naming Convention

1.4 Spatial Information

1.4.1 Coverage and Resolution

These data cover both the Northern and Southern polar regions at a 25 x 25 km grid cell size. Note: While resolution and grid cell size are often used interchangeably with regards to satellite data, there is an important difference. Resolution refers more properly to the instantaneous field of view (IFOV) of a particular sensor frequency. That is, resolution is the spot size on the ground that the sensor channel can resolve. The SSM/I channels used are the 19 GHz vertical, the 19 GHz horizontal, and the 37 GHz vertical. The IFOV of the 19 GHz SSM/I passive microwave channel is approximately 70 km x 45 km. See Table 2 in the C-ATBD (Meier et al., 2021) for a complete list of IFOVs by channel.

Since these data are gridded onto a 25 x 25 km grid and the IFOV of the sensor is coarser than this, the sensor is obtaining information from up to a 3×2 grid cell (~75 km \times 45 km) region, but because a simple drop-in-the-bucket gridding method is used, that signature is placed in a single grid cell. This results in a spatial "smearing" across several grid cells. Also, some grid cells do not coincide with the center of the sensor footprint and are thus left as missing even though there is brightness temperature information available at that region. Higher frequency channels have finer resolution, but because the sea ice concentration algorithms use data from the 19 GHz channel, the sea ice concentration estimate is affected by the makeup of the surface over an area considerably larger than the nominal 25 km resolution.

The spatial coordinates for the Northern polar region are the following:

Northernmost Latitude: 31.10° N Southernmost Latitude: 89.84° N Easternmost Longitude: 180° E Westernmost Longitude: 180° W

Note that for the Arctic, there is a region around the pole that is not imaged by the passive microwave sensors. This area is called the Arctic Pole Hole. Depending on the instrument used, the size of this area changes over time as the instrument changes. See [Table 9](#page-21-1) for these sizes.

With the release of Version 4, this area is now filled by spatial interpolation instead of being filled with missing values. Note, one cannot assume what the concentration is in the Arctic pole hole, especially in late Arctic summer and early autumn. Thus, we would advise caution in using the interpolated data in long-term trends or climatology analyses. See the C-ATBD (Meier et al., 2021) for more details.

Table 9. Arctic Pole Hole Size by Instrument

The spatial coordinates for the Southern polar region are the following:

Southernmost Latitude: 89.84° S Northernmost Latitude: 39.36° S Westernmost Longitude: 180° W Easternmost Longitude: 180° E

1.4.2 Projection and Grid Description

The sea ice concentration data are displayed in a polar stereographic projection. For more information on this projection, see the NSIDC [Polar Stereographic Projections and Grids](https://nsidc.org/data/polar-stereo/ps_grids.html) Web page. Note that the polar stereographic grid is not equal area; the latitude of true scale (tangent of the planar grid) is 70 degrees. Geolocation and grid details are given in [Table 10](#page-21-2) and [Table 11.](#page-22-1)

Table 11. Grid Details

1.5 Temporal Coverage and Resolution

The primary NOAA/NSIDC CDR sea ice concentrations (cdr_seaice_conc and cdr_seaice_conc_monthly) span 25 October 1978 to through most recent processing provided at both a daily resolution and a monthly averaged resolution [\(Table 12\)](#page-23-0). For the monthly averaged data, at least 15 days of data must be available for a month for an average to be calculated. There is a gap in the data from 03 December 1987 through 12 January 1988 due to satellite issues during that time, so no daily or monthly data are available for that time period. There are additional gaps in the data due to corrupt or missing data that are noted in [Table 13.](#page-23-1) Data files exist for these dates; however, they are filled with a missing data value of 255. In addition, dates of data that have partially corrupt data files are listed in [Table 14](#page-23-2) for reference, as they could cause issues in analyses of the time series because they contain values that look like sea ice concentration but that are clearly erroneous. Most of these data gaps occur during the SMMR era, which had some operational issues. See NSIDC Special Report 20 (Windnagel et al., 2021) for details on these corrupt and missing data. Many small gaps in the data (<10 days) are filled by temporal interpolation. See section [2.3.2.1](#page-26-0) [Temporal Gap Filling Notes](#page-26-0) for details.

In addition, a preliminary near-real-time version of this product is also available. The NRT SIC CDR is meant as an interim data set to fill the time period between updates of the final SIC CDR and to provide data up to the present. The NRT SIC CDR is preliminary and does not go through the

same quality control measures that the final SIC CDR does, so it should be treated as such. You can access this interim product here: *[Near-real-time NOAA/NSIDC Climate Data Record of Passive](https://nsidc.org/data/g10016) [Microwave Sea Ice Concentration](https://nsidc.org/data/g10016)*.

Platform and Instrument	Time Period
Nimbus-7 SMMR	25 October 1978 – 09 July 1987 Note: There are no data from 17 - 19 August 1984 due to satellite problems
DMSP-F8 SSM/L	10 July 1987 - 02 December 1991 Note: There are no data from 3 December 1987 through 12 January 1988 due to satellite problems.
DMSP-F11 SSM/I	03 December 1991 - 30 September 1995
DMSP-F13 SSM/L	01 October 1995 - 31 December 2007
DMSP-F17 SSMIS	01 January 2008 - most recent processing Note: There are no daily data for 24 March 2008, 25 March 2008, and 30 October 2008 due to satellite problems.

Table 12. Time Period Each Instrument is Used in the SIC CDR

Table 13. Daily and monthly dates with no data due to corrupt or missing data

Table 14. Dates of partial SIC CDR fields due to corrupt or missing data Note: Only dates where missing data affect sea ice concentration are noted here

2 DATA ACQUISITION AND PROCESSING

2.1 Input Data

The input data for the sea ice CDR variables are described in [Table 15.](#page-24-5)

2.2 Acquisition

The input gridded brightness temperatures used for creating the daily NOAA/NSIDC CDR sea ice concentrations (cdr_seaice_conc) are archived at NSIDC in two data sets listed in [Table 15.](#page-24-5) These SMMR gridded brightness temperatures were acquired from NASA and the SSM/I-SSMIS gridded brightness temperatures are produced by NSIDC from swath data obtained from [Remote](http://www.remss.com/) [Sensing Systems \(RSS\).](http://www.remss.com/) For a complete description of how the input data are processed, see the Data Acquisition and Processing sections in each data set user guide using the links in [Table 15.](#page-24-5) The input data for the monthly CDR concentration (cdr_seaice_conc_monthly) are the daily sea ice concentration CDR data.

2.3 Derivation Techniques and Algorithms

2.3.1 Overview

NSIDC processes the input brightness temperatures [\(Table 15\)](#page-24-5) into two different intermediate sea ice concentrations using two GSFC-developed algorithms: the NASA Team (NT) algorithm (Cavalieri et al., 1984) and the Bootstrap (BT) algorithm (Comiso, 1986). These intermediate NSIDC NT and BT sea ice concentrations are used in the NOAA/NSIDC SIC CDR algorithm described in further detail in the section [2.3.3](#page-28-0) SIC CDR [Algorithm.](#page-28-0)

The passive microwave channels employed for the sea ice concentration product are vertical (V) and horizontal (H) polarizations at 19 GHz (18.0 GHz for SMMR; 19.35 GHz for SSM/I and SSMIS), 22 GHz (V polarization only), and 37 GHz frequencies. [Table 16](#page-25-1) lists the channels used for each algorithm and the channels used for the weather filters. For a complete description of the weather filters, see the C-ATBD (Meier et al., 2021).

Since this data set uses multiple sensors over time, the sea ice algorithms are intercalibrated at the product (concentration) level by NASA GSFC. Thus, the brightness temperature source is less important because the intercalibration adjustment includes any necessary changes due to differences in brightness temperature across them. Both the NASA Team and Bootstrap algorithms employ varying tie-points to account for changes in sensors and spacecraft. These tie-point adjustments are derived from regressions of brightness temperatures during overlap periods. The adjustments are made at the product level by adjusting the algorithm coefficients so that the derived sea ice concentration fields are as consistent as possible.

The NASA Team approach uses sensor-specific hemispheric tie-points for each transition (Cavalieri et al., 1999; Cavalieri et al., 2011). Tie-points were originally derived for the SMMR sensor and subsequent transitions to the different SSM/I and SSMIS instruments adjusted the tiepoints to be consistent with the original SMMR record. The Bootstrap algorithm uses daily varying hemispheric tie-points, derived via linear regression analysis on clusters of brightness temperature values of the relevant channels (Comiso, 2009; Comiso and Nishio, 2008). Also, in contrast to the NASA Team, Bootstrap tie-points for SMMR, SSM/I, and SSMIS are derived from matching fields from the AMSR-E sensor, which is newer and more accurate.

2.3.2 Automated Quality Control

Automated quality control measures are implemented independently on the intermediate NASA Team and Bootstrap outputs. Two weather filters, based on ratios of channels sensitive to enhanced emission over open water, are used to filter weather effects. Separate land-spillover corrections are used for each of the algorithms to filter out much of the error due to mixed land/ocean grid cells. Finally, to screen out errant retrievals of ice in regions where sea ice never occurs, valid ice masks are applied to the Northern Hemisphere and climatological ocean masks are applied to the Southern Hemisphere. In addition, temporal and spatial gap filling have been

implemented for Version 4. For a complete description of the automated filters, masks, and gap filling, see the C-ATBD (Meier et al., 2021).

2.3.2.1 Temporal Gap Filling Notes

Gaps in the data can occur for many reasons from issues with the satellite, instrument, or ground stations collecting the data. Missing brightness temperature data can be in the form of no data at all for a day or more, entirely missing swath orbits, a few scans from a swath, or a few grid cells. To address these gaps and enhance the temporal and spatial completeness of the sea ice concentration CDR record, we have employed a temporal gap-filling approach described below along with guidelines for using the gap-filled data and an example of the effects of the method.

Two methods of temporal gap filling are performed on the data: two-sided and one-sided. The twosided method, attempted first, linearly interpolates missing data with weighted values from up to five days on either side of the missing date. These days do not have to be evenly spaced because the method is searching for the closest days to the missing date possible. For example, a missing grid cell can be interpolated from corresponding grid cells one day in the past and one day in the future if those data exist; or the method may have to search further into the past or future to find values to interpolate with, such as two days in the past and four days in the future. Once a past and future value is found, the method stops searching for a value to interpolate with. The interpolation is weighted, whereby data closer to the missing date (e.g. 1 day away) are given more weight than data further away (e.g. 5 days away). If data are not available within five days before or after a date, the one-sided method is applied. This simpler approach fills a missing grid cell with a copy of the data value from the closest corresponding grid cell from up to three days on either side of the date.

We chose five days for the two-sided interpolation and three days for the one-sided interpolation based on experience, though these choices were somewhat arbitrary. If neither method can be applied, the grid cell is marked as missing.

A flag called temporal interpolation flag marks the grid cells that were temporally interpolated. This flag uses one- or two-digit numbers to indicate the known data points used in the interpolation. For two-sided gap filling, it is always a 2-digit number where the first digit indicates the number of days in the past, while the second digit indicates the number of days in the future from which the data point came from, with a max of five days in either direction. For example, a flag value of 24 indicates that the missing grid cell was linearly interpolated using sea ice concentration data from two days prior and four days in the future. In the two-sided method, the flag values range from 11 to 55 but exclude 10, 20, and 30. For the one-sided gap filing, where only one day is used, the value can be one or two digits with possible values of 1, 2, 3, 10, 20, and 30. Two-digit values

indicate that data in the past were used, while single digit values indicate that data in the future were used. For example, a value of 30 indicates that data from three days in the past was copied.

Note: There is a bug in the one-sided interpolation whereby it is not looking for data into the future. Only the backward filling is being applied; so, values of 1, 2, or 3 can never occur. This bug will be fixed in a future version of the data product.

2.3.2.1.1 Guidelines for Using Temporally Interpolated Data

The temporal interpolation flag is provided as a way for users to screen for temporally gap-filled data. While the interpolation aims to provide the most complete fields possible, users can decide how much (if any) interpolation they wish to use based on the flag values. Here are some guidelines to consider when using the temporally interpolated data:

- The farther away from the day in question (i.e., the longer time period one is interpolating across) the less reliable the estimate.
- An asymmetry in the interpolation can also make the estimate less reliable (e.g. using data from 1 day in the past and 3 days in the future). Sea ice tends to grow linearly, so symmetrically weighted interpolation (i.e., same size gap before and after) typically yields reasonably good results. However, asymmetric, or especially one-directional, interpolation is less reliable.
- Another aspect is what spatial scale one is looking at. If one is looking at total extent or area for the entire Arctic or Antarctic, there is less sensitivity to interpolation because effects will average out. But if one is looking at a smaller region, then the interpolation could produce some odd-looking results.

2.3.2.1.2 Temporal Interpolation Example

Below is one example of how temporal gap filling works. In this example, two-sided and one-sided temporal gap filling work in conjunction. The example also illustrates the consequences of the bug in the one-sided gap filling noted above.

Due to an issue with the DMSP F17 satellite, there is no data at all for seven days from 19 March to 25 March 2008. The code attempts to fill this with temporal interpolation. Because this gap is larger than five days, a mix of the two-sided and one-sided interpolation is applied. The following bullets and [Figure 1](#page-28-1) describe what occurs for each grid cell and for each day during this gap:

- **March 18** is real data so the temporal interpolation flag is 0, i.e. no interpolation.
- **March 19** is a copy of March 18 because there is no data within five days into the future, so the one-sided gap filling technique is used. The temporal_interpolation_flag value is 10 indicating that the missing grid cells were filled with a copy of the data from one day prior.
- **March 20** is also a copy of March 18 because there is no data within five days into the future, so the one-sided gap filling technique is used. The temporal_interpolation flag

value is 20, indicating that the missing grid cells were filled with the data from two days prior.

- **March 21** does have data within five days on either side, so it is gap filled using the twosided method. The temporal interpolation flag values for this data are 35, so it is linearly interpolated with data from 3 days prior (March 18) and 5 days in the future (March 26).
- **March 22** does have data within five days on either side, so it is gap filled using the twosided method. The temporal interpolation flag values for this data are 44, so it is linearly interpolated with data from 4 days prior (March 18) and 4 days in the future (March 26).
- **March 23** does have data within five days on either side, so it is gap filled using the twosided method. The temporal_interpolation_flag values for this data are 53, so it is linearly interpolated with data from 5 days prior (March 18) and 3 days in the future (March 26).
- **March 24** is filled with missing but should be filled with a copy of data from March 26, since it's two days in the future, which is less than the three-day limit. This illustrates the bug in the one-sided method noted above.
- **March 25** is filled with missing but should be filled with a copy of data from March 26, since it's two days in the future, which is less than the three-day limit. This illustrates the bug in the one-sided method noted above.

Note that there is a jump from March 20 to March 21 since March 20 is a copy of March 18, but March 21 is linearly interpolated from data on March 18 and March 26.

Figure 1. Temporal gap filling for the 7-day gap from 19 - 25 March 2008.

2.3.3 SIC CDR Algorithm

Different algorithms exist for computing sea ice concentration from brightness temperature data. The two widely used GSFC-developed NASA Team (Cavalieri et al., 1984) and Bootstrap (Comiso, 1986) algorithms are described in sections [2.3.4](#page-29-0) an[d 2.3.5,](#page-31-0) respectively. Both algorithms have their own inherent advantages and limitations. For this SIC CDR data set, NSIDC processes the input brightness temperatures into two intermediate NASA Team and Bootstrap sea ice concentration fields following the way NASA produces their NASA Team and Bootstrap data sets with a few small differences. See the Theoretical Description section of the C-ATBD (Meier et al., 2021) for full details. The NASA-produced products are available from NSIDC as the *[Sea Ice Concentrations](http://nsidc.org/data/nsidc-0051) [from Nimbus-7 SMMR and DMSP SSM/I-SSMIS Passive Microwave Data](http://nsidc.org/data/nsidc-0051)* and the *[Bootstrap Sea](https://nsidc.org/data/nsidc-0079) [Ice Concentrations from Nimbus-7 SMMR and DMSP SSM/I-SSMIS](https://nsidc.org/data/nsidc-0079)*.

The NASA Team-derived sea ice concentrations are then merged with the Bootstrap-derived sea ice concentrations into a single ice concentration estimate. The SIC CDR algorithm steps are as follows:

- The sea ice concentrations estimated by the Bootstrap algorithm are analyzed first. Any grid cell with a concentration estimate of 10% or greater will be considered valid ice in the final product.
- For each grid cell that passes the Bootstrap threshold, the concentration value given by the NASA Team algorithm is compared with that given by the Bootstrap algorithm; whichever value is greater is selected as the CDR value.
- Any concentration values higher than 100% are set to 100%.

The resulting SIC CDR field has sea ice concentration values as low as 10% and as high as 100%.

Numerous studies, some noted below, have shown that passive microwave-based algorithms tend to underestimate true ice concentration. While both NASA Team and Bootstrap algorithms underestimate, the NASA Team algorithm tends to underestimate by a greater amount. The basis of the SIC CDR algorithm is that when compared, the algorithm estimate that is the highest concentration for a given grid cell is likely to be the more accurate estimate. The Bootstrap algorithm runs first because it generally does a better job at detecting ice in areas of low concentration and where the ice is thin.

The NASA Team algorithm, because it uses a ratio of brightness temperatures, tends to cancel out any physical temperature effects. The Bootstrap algorithm uses relationships between two brightness temperatures that are dependent on physical temperature. Thus, physical temperature changes can affect Bootstrap estimates. Errors occur primarily in regimes with very low temperatures: winter in the high Arctic and near the Antarctic coast (Comiso et al., 1997), where the Bootstrap algorithm can underestimate concentration and give a lower value than the NASA Team algorithm. During winter conditions with more moderate temperatures, NASA Team concentrations also tend to have more of a low bias (Kwok, 2002; Meier, 2005). During melt conditions, both algorithms tend to underestimate concentration; but the effect is more pronounced in the NASA Team algorithm (Comiso et al., 1997; Meier, 2005; Andersen et al., 2007).

While these characteristics of the algorithm are true in an overall general sense, ice conditions and algorithm performance can vary from grid cell to grid cell; and in some cases, this approach of choosing the larger value will result in an overestimation of concentration (Meier, 2005). However, using the higher concentration between the two algorithms will tend to reduce the overall underestimation of the SIC CDR estimate (Meier et al., 2014). For a more in-depth discussion on the reasoning behind the algorithm, see the Theoretical Description section of the C-ATBD (Meier et al., 2021).

2.3.4 NASA Team Algorithm

The NASA Team algorithm uses brightness temperatures from the 19 GHz V, 19 GHz H, and 37 GHz V channels. The methodology is based on two brightness temperature ratios, the polarization ratio (PR) of the 19 GHz V and H channels (Equation 1) and the spectral gradient ratio (GR) of the 19 GHz V and 37 GHz V channels (Equation 2).

$$
PR(19) = [TB(19V) - TB(19H))] / [TB(19V) + TB(19H)]
$$
 (Equation 1)
GR(37V/19V) = [T_B(37V) - T_B(19V)] / [T_B(37V) + T_B(19V)] (Equation 2)

Where:

Table 17. NASA Team Algorithm Variable Descriptions

When PR and GR are plotted against each other, brightness temperature values tend to cluster in two locations, an open water (zero percent ice) point and a line representing 100 percent ice concentration, roughly forming a triangle. The concentration of a grid cell with a given GR and PR value is calculated by a linear interpolation between the open water point and the 100 percent line segment. See [Figure 2.](#page-31-1)

For a detailed description of the NASA Team algorithm, please see the [Descriptions of and](https://nsidc.org/support/faq/nasa-team-vs-bootstrap-algorithm) [Differences Between the NASA Team and Bootstrap Algorithms FAQ](https://nsidc.org/support/faq/nasa-team-vs-bootstrap-algorithm) and the [NASA Technical](https://nsidc.org/sites/nsidc.org/files/technical-references/NASA%20Technical%20Memorandum%20104647.pdf) [Memorandum 104647](https://nsidc.org/sites/nsidc.org/files/technical-references/NASA%20Technical%20Memorandum%20104647.pdf) (Cavalieri et al., 1997) that includes information about differences (for example, tie points) between the original algorithm and the revised NASA Team algorithm, and the NASA Team Algorithm section of the C-ATBD (Meier et al., 2021) for a table of tie-point values.

Figure 2. Sample plot of GR vs. PR with typical clustering of grid cell values (small dots) around the 0% ice (open water) point (blue star) and the 100% ice line (circled in red). Points with a mixture of ice and water (circled in green) fall between these two extremes. Adapted from Figure 10-2 of Steffen et al. (1992).

2.3.5 Bootstrap Algorithm

Like the NASA Team algorithm, the Bootstrap algorithm is empirically derived based on relationships of brightness temperatures at different channels. The Bootstrap method uses the fact that scatter plots of different sets of channels show distinct clusters that correspond to two pure surface types: 100 percent sea ice or open water.

[Figure 3](#page-32-0) shows a schematic of the general relationship between two channels. Points that fall along line segment AD represent 100 percent ice cover. Points that cluster around point O represent open water (zero percent ice). Concentration for a point B is determined by a linear interpolation along the distance from O to I where I is the intersection of segment OB and segment AD. This is described by Equation 3.

$$
C = (T_B - T_O)/(T_I - T_O)
$$

Where:

Table 18. Bootstrap Algorithm Variable Descriptions

Variable	Description
	Sea ice concentration
Τв	Observed brightness temperature
Τo	Reference brightness temperatures for open water
	Reference brightness temperatures for sea ice

Figure 3. Example of the relationship of the 19V vs. 37V T_B (in Kelvin) used in the Bootstrap algorithm. Brightness temperatures typically cluster around the line segments AD (representing 100% sea ice) and OW (representing 100% open water). For points that fall below the AD-5 line (dotted line), Bootstrap uses T_B relationships for 37H vs. 37V. Adapted from Comiso and Nishio (2008).

The Bootstrap algorithm uses two such combinations, 37 GHz H versus 37 GHz V and 19 GHz V versus 37 GHz V, denoted as HV37 and V1937, respectively. Points that fall within 5 K of the AD segment in a HV37 plot, corresponding roughly to concentrations greater than 90 percent, use this approach. Points that fall below the AD-5 line, use the V1937 relationship to derive the concentration. Slope and offset values for line segment AD were originally derived for each hemisphere for different seasonal conditions (Table 2 in Comiso et al., 1997). However, a newer formulation, employed in this SIC CDR, was developed where slope and offsets are derived for

each daily field based on the clustering within the daily brightness temperatures (Comiso and Nishio, 2008). For a detailed description of the Bootstrap algorithm, please see the [Descriptions of](https://nsidc.org/support/faq/nasa-team-vs-bootstrap-algorithm) [and Differences Between the NASA Team and Bootstrap Algorithms FAQ.](https://nsidc.org/support/faq/nasa-team-vs-bootstrap-algorithm)

2.4 Processing Steps

Below are the processing steps for both the daily and monthly data files. In addition, the source code is provided for transparency of the algorithm and processes used in creating the sea ice CDR. This source code is for reference only and is not intended to be portable to any computer system beyond that of the original SIC CDR producer's environment. You can access the code from the NOAA Climate Data Record Program's Operation CDR Web page under the Oceanic CDR[s Sea](https://www.ncdc.noaa.gov/cdr/oceanic/sea-ice-concentration) [Ice Concentration](https://www.ncdc.noaa.gov/cdr/oceanic/sea-ice-concentration) section.

2.4.1 Daily Files

The following are the general steps NSIDC uses to produce the daily NOAA/NSIDC CDR sea ice concentration product. Se[e Figure 4](#page-34-1) for a diagram of the data flow.

- 1. Obtain input brightness temperatures from the NSIDC *[Nimbus-7 SMMR Polar Gridded](https://nsidc.org/data/nsidc-0007) [Radiances and Sea Ice Concentrations](https://nsidc.org/data/nsidc-0007)* (NSIDC-0007) data set and the *[DMSP SSM/I-](http://nsidc.org/data/nsidc-0001)[SSMIS Daily Polar Gridded Brightness Temperatures](http://nsidc.org/data/nsidc-0001)* (NSIDC-0001) data set. See [Table](#page-25-1) [16](#page-25-1) for a list of channels used.
- 2. Spatially interpolate each brightness temperature channel. Fill the spatial interpolation flag variable. See the Quality Control Procedures section of the C-ATBD (Meier et al., 2021) for details.
- 3. Process the brightness temperatures into two intermediate sea ice concentration products using both the NASA Team and Bootstrap algorithms.
- 4. Apply weather filters, land-spillover corrections, and monthly valid ice masks.
- 5. Set some initial QA flags (qa_of_cdr_seaice_conc) based on the filters in step 4.
- 6. Temporally interpolate the intermediate NASA Team and Bootstrap sea ice concentrations. See the Quality Control Procedures section of the C-ATBD (Meier et al., 2021) for details.
- 7. For the Arctic, spatially interpolate the pole hole. See the Quality Control Procedures section of the C-ATBD (Meier et al., 2021) for details.
- 8. Merge the intermediate NSIDC NASA Team (nsidc_nt_seaice_conc) and Bootstrap (nsidc_bt_seaice_conc) data into the final SIC CDR using the SIC CDR algorithm and populate the cdr_seaice_conc_variable. See section [2.3.3](#page-28-0) SIC CDR [Algorithm](#page-28-0) of this document for more information.
- 9. Apply a final weather filter and land-spillover correction and apply a day-of-year valid ice mask for the SMMR era to the sea ice concentration CDR.
- 10. Compute the CDR sea ice concentration standard deviation (stdev of cdr seaice conc) and the final QA flag values (qa of cdr seaice conc).
- 11. Calculate melt onset (melt_onset_day_cdr_seaice_conc) and add melt-indicator flag to the QA variable (qa_of_cdr_seaice_conc) via a post-processing step.
- 12. Populate the daily NetCDF variables and create the .nc files.

Figure 4. Flow of Data through the Daily SIC CDR Processing.

2.4.2 Monthly Files

The following are the general steps NSIDC uses to produce the monthly NOAA/NSIDC CDR sea ice concentration product. See [Figure 5](#page-35-1) for a diagram of the data flow.

- 1. Read the input daily NSIDC NASA Team and Bootstrap sea ice concentration data (nsidc nt seaice conc and nsidc bt seaice conc).
- 2. Compute the monthly mean concentration for each grid cell for a given month from the daily NASA Team and Bootstrap values.
- 3. Merge the monthly intermediate NSIDC NASA Team (nsidc_nt_seaice_conc_monthly) and Bootstrap (nsidc_bt_seaice_conc_monthly) data into the final SIC CDR using the SIC CDR algorithm and populate the cdr_seaice_conc_monthly variable. See section [2.3.3](#page-28-0) SIC CDR [Algorithm](#page-28-0) of this document for more information.
- 4. Compute the standard deviation and quality flags and fill those variables (stdev_of_cdr_seaice_conc_monthly and qa_of_cdr_seaice_conc_monthly).
- 5. Set melt onset day (value from the last day of the month) and fill the melt onset day cdr seaice conc monthly variable and add melt onset flag to the qa of cdr seaice conc monthly variable.
- 6. Populate the monthly NetCDF variables and create the .nc files.

Figure 5. Flow of Data through the Monthly SIC CDR Processing.

2.5 Errors Sources

Several studies over the years have assessed sea ice concentration estimates from the NASA Team and Bootstrap algorithms. These assessments have typically used coincident airborne or satellite remote sensing data from optical, thermal, or radar sensors, generally at a higher spatial resolution than the SSM/I instrument but with only local or regional coverage. Several assessments indicate an accuracy of approximately five percent during mid-winter conditions away from the coast and the ice edge (Comiso et al., 1997; Ivanova et al., 2015; Kern et al., 2019; Meier et al., 2005; Andersen et al., 2007). Other assessments suggest concentration estimates are less accurate. Kwok (2002) found that passive microwave overestimates open water by three to five times in winter. Partington et al. (2003) found a difference with operational charts that was relatively low in the winter but rose to more than 20 percent in summer. In summer, due to surface melt and melt ponding, errors are higher (e.g., Kern et al., 2016; Kern et al., 2020).

Researchers can assess and improve a CDR by comparing it with operational products — real-time products that help ships cross the sea ice. Absolute error can be approximated via comparison to operational sea ice products, such as those produced by the [U.S. National Ice Center](https://usicecenter.gov/) (USNIC) or the [Canadian Ice Service,](https://www.canada.ca/en/environment-climate-change/services/ice-forecasts-observations/latest-conditions.html) but it is important to keep in mind that such products have an operational

focus different from the climate focus of the SIC CDR, and the two are not expected to be consistent with each other. The documentation for the daily *[Multi-Sensor Analyzed Sea Ice](http://nsidc.org/data/masie) [Extent](http://nsidc.org/data/masie)* (MASIE), distributed by NSIDC in cooperation with USNIC, gives a summary of how satellite passive microwave SIC CDRs differ from operational products. An example of how the 15% contour in this SIC CDR product typically differs from the ide edge that analysts would find is given in Fetterer et al. (2021).

Errors can come from problems with the sensor, from weather effects, and from inadequacies in the algorithm. A satellite's orbit may drift over time, for example, which may degrade the data quality of an instrument. Most SSM/I instruments were used long past their designed lifetime expectancy. Atmospheric water vapor is a weather effect that can modulate the passive microwave signature of the surface, particularly at the 19 GHz frequency, causing ice concentration to be overestimated. The emissivity of sea water is generally stable, except under strong winds that cause waves to form. The emissivity of sea ice varies considerably depending on many factors including age, thickness, and surface roughness. When one considers that algorithms must arrive at a single number for ice concentration, taking into account the varying brightness temperatures of all the different surface types that may fill the footprints of the 19 GHz and 37 GHz channels, and that those footprints differ in size and shape across the instrument swath, one can appreciate the difficulty of the problem. Microwave Remote Sensing of Sea Ice, edited F. Carsey, provides a comprehensive overview of the subject (Carsey, 1992).

Another potential sensor error results from the transition between sensors on different platforms. The brightness temperature regression and tie-point adjustment corrects for this, though small artifacts remain (Cavalieri et al., 1999; Comiso and Nishio, 2008). Comparison of ice extent estimates from sensor overlap periods indicate that the adjustments yield agreements that are on the order of 0.05 percent or less and about 0.5 percent for sea ice area (Cavalieri et al., 1999; Cavalieri et al., 2011). Short overlap periods of early sensor transitions (SMMR to F8 and F8 to F11) may not account for the full seasonal variability (Meier and Khalsa, 2011b; Cavalieri et al., 2011) and differences may be higher in some cases. However, differences appear to be well below the sensitivity of the instrument, thus, providing confidence in the robustness of the intercalibrated algorithms through the time series.

When melt ponds form on the surface of ice floes in the summer, the ice concentration appears to decline when in fact the true concentration may not have changed (Fetterer and Untersteiner, 1998). Melt state is a surface effect that may in itself contain a climate trend, which could influence sea ice concentration trend estimates. This and other concentration error sources have been examined to some extent in Andersen et al. (2007), and their influence appears to be small compared to the estimated sea ice trends, but such effects should be kept in mind when using these data.

The NetCDF4 files contain a variable called qa_of_cdr_seaice_conc to help data users assess the quality of a given data value. For example, if the melt onset value is set, then there is higher error in the sea ice concentration value at that grid cell. [Table 3](#page-7-2) gives a list of the flags, their values, and their meaning. **Note**: Grid cells that meet multiple conditions will have a value that is the sum of the values of each individual condition.

For a more complete description of error sources and assessments, see the C-ATBD (Meier et al., 2021).

2.6 NSIDC-Processed NASA Team and Bootstrap Data Versus the GSFC-Produced Data

The NSIDC-processed NASA Team and Bootstrap sea ice concentration variables (nsidc_nt_seaice_conc and nsidc_bt_seaice_conc) are very similar to the GSFC-produced concentration available from NSIDC as the *[Sea Ice Concentrations from Nimbus-7 SMMR and](http://nsidc.org/data/nsidc-0051) [DMSP SSM/I-SSMIS Passive Microwave Data](http://nsidc.org/data/nsidc-0051)* and the *[Bootstrap Sea Ice Concentrations from](https://nsidc.org/data/nsidc-0079) [Nimbus-7 SMMR and DMSP SSM/I-SSMIS](https://nsidc.org/data/nsidc-0079)*. Although the differences are subtle, they are important.

The GSFC-produced concentrations include thorough quality control, including manual correction/replacement of bad values (for example, false ice due to weather effects over the ocean), and spatial or temporal interpolation to fill in missing values.

The NSIDC-processed concentrations are based on the same NASA Team and Bootstrap algorithms at Goddard but run in-house at NSIDC. There is automated spatial/temporal interpolation applied but there is no manual quality control.

The NSIDC-processed concentrations were created to meet the [NOAA CDR Program](https://www.ncei.noaa.gov/products/climate-data-records) criteria most notably fully transparent and reproducible processing. The GSFC data do not meet this requirement because of the manual quality control aspect. However, this manual quality control likely results in a more accurate product but the automated quality control applied to the NSIDC concentrations is a good approximation of the manual inspection and is fully traceable.

For more details on the differences between the two different concentration products, see Meier et al. (2014) and Peng et al. (2013) which compare the two and show that the differences at the total extent/area level are mostly in the noise and there is very little difference in terms of trends.

2.7 Instrumentation

For the NOAA/NSIDC SIC CDR data, NSIDC uses brightness temperatures from the SMMR sensor on Nimbus-7 satellite, SSM/I sensors on the DMSP-F8, -F11, and -F13 platforms, and from the SSMIS sensor on DMSP-F17. See [Table 19](#page-38-3) for a description of orbital parameters of the different platforms. The rationale for using only these satellites was made to keep the equatorial crossing times as consistent as possible to minimize potential diurnal effects of data from sunsynchronous orbits of the DMSP satellites. For a list of the footprint size of each sensor by channel, see Table 2 in the C-ATBD (Meier et al., 2021).

*Indicates sensor and spacecraft orbital characteristics of the three sensors used in generating the sea ice concentrations.

3 SOFTWARE AND TOOLS

3.1 NetCDF Description

The NetCDF format is a self-describing file format for array-oriented data. For more information on this file format see the NSIDC [What is NetCDF?](https://nsidc.org/data/user-resources/help-center/what-netcdf) web page and the links therein.

3.2 STAC Catalog

We provide a [SpatioTemporal Asset Catalog](https://stacspec.org/en) (STAC) to improve the accessibility and discoverability of the SIC CDR. A STAC catalog provides special metadata about the collection that provides a standardized way to expose the spatio-temporal data and allows it to be more easily worked with, indexed, and discovered. In this section, we describe the catalog and provide examples of how to utilize it.

A STAC catalog uses a JSON schema to describe the data via different STAC specifications. For this STAC, we use the *Collection* and *Item* specifications. The *Collection* specification is used to describe a group of related *Items* using a set of common metadata fields. In this case, the *Collection* is the entire SIC CDR data product. The *Item* specification is used to describe each asset of the *Collection*. In this case, each individual SIC CDR NetCDF file. Thes two specifications provide product-level (*Collection*) and file-level (*Item*) metadata for the data set.

This catalog begins with a JSON file that defines the *Collection* in a file called collection.json. Each *Item* represents a single spatio-temporal asset of the catalog, and each is described in its own JSON file. For the SIC CDR, there is one *Item* file for each daily and monthly aggregated NetCDF file for the northern and southern hemispheres. The *Item* JSON files have the following file naming conventions:

Daily aggregated: seaice_conc_daily_[Xh]_[YYYY]_v04r00.json

Monthly aggregated: seaice_conc_monthly_[Xh]_197811_[YYYYMM]_v04r00.json

Where:

This STAC Catalog is located here: [https://noaadata.apps.nsidc.org/NOAA/G02202_V4/stac/.](https://noaadata.apps.nsidc.org/NOAA/G02202_V4/stac/)

Examples of how to utilize the STAC catalog using Python can be found on NSIDC's GitHub repository in the Sea Ice CDR Learning Notebooks tutorial at [https://github.com/nsidc/seaice-cdr](https://github.com/nsidc/seaice-cdr-learning-notebook)[learning-notebook.](https://github.com/nsidc/seaice-cdr-learning-notebook)

4 VERSION HISTORY

Table 20. Version History

5 RELATED DATA SETS

- *[Near-real-time NOAA/NSIDC Climate Data Record of Passive Microwave Sea Ice](https://nsidc.org/data/g10016) [Concentration](https://nsidc.org/data/g10016)*
- *[DMSP SSM/I-SSMIS Daily Polar Gridded Brightness Temperatures](http://nsidc.org/data/nsidc-0001)*
- *[Sea Ice Concentrations from Nimbus-7 SMMR and DMSP SSM/I Passive Microwave Data](http://nsidc.org/data/nsidc-0051)*
- *[Bootstrap Sea Ice Concentrations from Nimbus-7 SMMR and DMSP SSM/I](http://nsidc.org/data/nsidc-0079)*
- *[Multi-sensor Analyzed Sea Ice Extent \(MASIE\)](http://nsidc.org/data/masie)*
- *[Sea Ice Index](http://nsidc.org/data/seaice_index)*
- *[Gridded Monthly Sea Ice Extent and Concentration, 1850 Onward](https://nsidc.org/data/g10010)*
- *[AMSR-E/Aqua Daily L3 12.5 km Brightness Temperatures, Sea Ice Concentration, & Snow](http://nsidc.org/data/ae_si12) [Depth Polar Grids](http://nsidc.org/data/ae_si12)*
- *[AMSR-E/Aqua Daily L3 25 km Brightness Temperatures & Sea Ice Concentration Polar Grids](http://nsidc.org/data/ae_si25)*

6 RELATED WEBSITES

- [NOAA's National Centers for Environmental Information \(NCEI\)](https://www.ncei.noaa.gov/products/climate-data-records) Climate Data Record (CDR) [program](https://www.ncei.noaa.gov/products/climate-data-records)
- [EUMETSAT Ocean & Sea Ice Satellite Application Facility](https://www.eumetsat.int/osi-saf)
- [Sea Ice Concentration: NOAA/NSIDC Climate Data Record:](https://climatedataguide.ucar.edu/climate-data/sea-ice-concentration-noaansidc-climate-data-record) Provides an overview of the data product's strengths and weaknesses (Meier and NCAR, 2014).

7 CONTACTS AND ACKNOWLEDGMENTS

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9 DOCUMENT INFORMATION

9.1 Author

A. Windnagel

9.2 Publication Date

July 2011

9.3 Revision History

August 2024: A. Windnagel updated the document to describe the addition of the STAC catalog and added more information on how the temporal interpolation is performed.

May 2021: A. Windnagel updated the document to reflect changes with the release of Version 4 Revision 0.

October 2018: A. Windnagel updated the version history section to note the release of the 2017 data and added a technical note about the Bootstrap data to the Input Data section.

December 2017: A. Windnagel updated the version history section to note the changes and updates to Version 3 Revision 1.

August 2017: A. Windnagel updated the document to represent Version 3 Revision 0 changes and updates.

May 2016: A. Windnagel updated the document with the Variables at a Glance tables and made other minor edits.

August 2015: A. Windnagel updated the flow chart diagrams and the version history to reflect the new modularization done to the code.

June 2015: A. Windnagel added the Differences in the NOAA/NSIDC Concentration CDR Variables and the Merged GSFC-Produced Concentration Variables section to clarify which variable to use.

July 2014: A. Windnagel updated the temporal coverage to reflect the new 2013 data that was processed.

March 2013: A. Windnagel updated the document to describe the new Version 2 Revision 00 of these data. Added new processing flowcharts, new melt variable description, and updated the description of the melt detection QA flag. Also added that the temporal coverage now spans through 2012.

May 2012: A. Windnagel added the monthly file information and put the document into the new guide doc style.