Coastal Ocean Data Analysis Product in North America (CODAP-NA) - An internally consistent data product for discrete inorganic carbon, oxygen, and nutrients on the U.S. North American ocean margins

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Abstract. Internally-consistent, quality-controlled data products play a very important role in promoting regional to global research efforts to understand societal vulnerabilities to ocean acidification (OA). However, there are currently no such data products for the coastal ocean where most of the OA-susceptible commercial and recreational fisheries and aquaculture industries are located. In this collaborative effort, we compiled, quality controlled (QC), and synthesized two decades of

40 discrete measurements of inorganic carbon system parameters, oxygen, and nutrient chemistry data from the U.S. North American continental shelves, to generate a data product called the Coastal Ocean Data Analysis Product for North America (CODAP-NA). There are few deep-water (>1500m) sampling locations in the current data product. As a result, cross-over analyses, which rely on comparisons between measurements on different cruises in the stable deep ocean, could not form the basis for cruise-to-cruise adjustments. For this reason, care was taken in the selection of data sets to include in this initial

45 release of CODAP-NA, and only data sets from laboratories with known quality assurance practices were included. New consistency checks and outlier detections were used to QC the data. Future releases of this CODAP-NA product will use this core data product as the basis for secondary QC. We worked closely with the investigators who collected and measured these data during the QC process. This version of the CODAP-NA is comprised of 3,292 oceanographic profiles from 61 research cruises covering all continental shelves of North America, from Alaska to Mexico in the west and from Canada to the

50 Caribbean in the east. Data for 14 variables (temperature; salinity; dissolved oxygen concentration; dissolved inorganic carbon concentration; total alkalinity; pH on the Total Scale; carbonate ion concentration; fugacity of carbon dioxide; and concentrations of silicate, phosphate, nitrate, nitrite, nitrate plus nitrite, and ammonium) have been subjected to extensive QC. CODAP-NA is available as a merged data product (Excel, CSV, MATLAB, and NetCDF, doi:10.25921/531n-c230, https://www.ncei.noaa.gov/data/oceans/ncei/ocads/metadata/0219960.html) (Jiang et al., 2020). The original cruise data have

55 also been updated with data providers' consent and summarized in a table with links to NOAA's National Centers for Environmental Information (NCEI) archives (https://www.ncei.noaa.gov/access/ocean-acidification-data-stewardship-oads/synthesis/NAcruises.html).

1 Introduction

Anthropogenic ocean acidification (OA) refers to the process by which the ocean's uptake of excess anthropogenic

60 atmospheric carbon dioxide (CO_2) reduces ocean pH and calcium carbonate mineral saturation states (Feely et al., 2004; Orr et al., 2005; Jiang et al., 2019; IPCC, 2011). OA is making it more difficult for marine calcifiers to build a shell and/or skeletal structure, endangering coral reefs and other marine ecosystems (Doney et al., 2009; Gattuso and Hanson, 2011). Despite only covering ~20% of Earth's land surface, coastal regions (from the coastline up to 200 km inland) host over 50% of the entire human population (Small and Nicholls, 2003; Hugo, 2011; Neumann et al., 2015). Coastal ecosystems account

65 for most of the economic activities related to commercial and recreational fisheries and aquaculture industries, supporting

about 90% of the global fisheries yield and 80% of known species of marine fish (Cicin-Sain et al., 2002). Studies have shown that OA has the potential to significantly impact both the fisheries and aquaculture industries, and change the way humans make their living, run their communities, and live their lives in coastal regions around the world (Cooley and Doney, 2009; Barton et al., 2012, 2015).

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The Global Ocean Data Analysis Project (GLODAPv2) offers an internally consistent data product for discrete sampling-based, open-ocean carbonate chemistry, nutrient chemistry, isotopes, and transient tracer data (Olsen et al., 2016; Olsen et al., 2020), allowing for a slew of new research products related to OA and its temporal trends in the global ocean (e.g., Lauvset et al., 2015; Jiang et al., 2015a; Gruber et al., 2019; Lauvset et al., 2020). While there are several coastal surface water partial

75 pressure of CO_2 (pCO_2) data products and climatologies (e.g., Bakker et al., 2016; Laruelle et al., 2017; Roobaert et al., 2019; Takahashi et al., 2020), internally consistent data products for water column carbonate and nutrient chemistry data in the coastal ocean currently do not exist. Such products would contribute significantly to our understanding of the current status of OA and its temporal trends, and help guide OA mitigation and adaptation efforts in coastal oceans where most of the global fisheries and aquaculture industries are focused.

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The impact of OA on North American ocean margins is expected to vary significantly from region to region, with distinct regional drivers amplifying or mitigating overall coastal acidification. Anthropogenic carbon dioxide (CO_2) invasion has been identified as the primary driver of open ocean acidification over decadal time scales, but coastal ocean acidification is influenced by many other physical, biological, and anthropogenic processes that can oppose or amplify the anthropogenic

85 CO_2 uptake. The U.S. continental West Coast (WC) and East Coast (EC) are in two vastly different ocean basins (Pacific vs. Atlantic) with different amounts of net organic matter remineralization in deeper waters flowing along the path of the Global Thermohaline Circulation (Broecker, 1991; Feely et al., 2008; Jiang et al., 2010; Wanninkhof et al., 2015). In the surface ocean, latitudinal variation of sea surface temperature (SST) and the ratio of dissolved inorganic carbon (DIC) to total alkalinity (TA) result in significantly different pH and calcium carbonate mineral saturation states between the Alaska Coast

90 and Gulf of Mexico (Jiang et al., 2019; Cai et al., 2020). Upwelling can bring deep waters with corrosive OA chemistry (resulting from large respiratory CO_2 loads) to the surface, while onshore surface flow can bring less-corrosive open ocean waters to the coastline (Hales et al., 2005; Feely et al., 2008, 2016). Riverine input of low-pH water is found to intensify OA shoreward of the shelf break on the EC (Hunt et al., 2011; Xue et al., 2016). However, riverine water composition also varies significantly and the Mississippi River is a source of high-TA water to the Gulf of Mexico (Cai et al., 2008; Stets et al., 2014;

95 Gomez et al., 2020). Eutrophication (enhancement of biological production of organic matter through addition of nutrients) causes high pH and calcium carbonate mineral saturation states in surface waters of the coastal ocean, and can lead to subsurface hypoxia (via subsequent respiration of that production), which is associated with low pH and calcium carbonate mineral saturation (Borges and Gypens, 2010; Cai et al., 2011; Laurent el al., 2017; Feely et al., 2016, 2018). The lack of OA

synthesis efforts on North American ocean margins hampers our understanding of the geographic pattern and relative

100 regional progression rates of OA along these coastlines (Cai et al., 2020).

Carbonate data in the coastal ocean are often collected by multiple laboratories with different methods and instruments. Many of the data sets may have never been shared with any major data centers, nor have these data sets gone through rigorous quality control (QC) and inter-comparison analyses. The lack of observations in intermediate and deep water (water

105 depth >1500 m) makes it challenging to adjust the data based on constancy of parameters in deep water (i.e., cross-over analyses) as is done for the open ocean (Lauvset and Tanhua, 2015). All these factors contribute to the lack of internally consistent data products for these important coastal environments. In this study, we compiled and QCed discrete sampling-based data for inorganic carbon, oxygen, and nutrient chemistry, and hydrographic parameters collected from the entire U.S. North American continental shelves. We serve both the internally-consistent climate quality data product, as well as the

110 QCed original cruise data through the NOAA National Centers for Environmental Information (NCEI). This effort will promote future OA research, modeling, and data synthesis in critically important coastal regions to help advance the OA adaptation, mitigation, and planning efforts of U.S. coastal communities. While we only partially address these limitations in this study, we do produce a data product that can be used as the basis to address these limitations and incorporate additional coastal cruises going forward. We hope this release will be considered analogous to GLODAPv2 (Olsen et al. 2016), in the

115 sense that the new data sets added in the subsequent GLODAPv2.2019 and .2020 updates (Olsen et al., 2019; 2020) were brought to be internally consistent with the fully quality-controlled data in the original GLODAPv2 product.

2 Study area

From a geopolitical perspective, the term "continental shelf" is defined as the region between the coastline (excluding estuaries) and a distance of 200 nautical miles (~370 km) offshore. While this definition is not as mechanistic as one based

120 on a change in bathymetric gradient or a hydrographic condition such as chlorophyll or salinity levels, it is regionally and seasonally invariant, and captures the full extent of coastal influences (Hales et al., 2008). This version of the data product is focused on the continental shelves of the U.S. North American (NA) coasts (Figure 1), including:

- Alaska Coast (AC) – covering the large marine ecosystems (LMEs) of Gulf of Alaska, East Bering Sea, Northern Bering-Chukchi Seas, and Beaufort Sea (see Sherman et al., 2009 for more information on the LMEs).
125 - U.S. West Coast (WC) – covering the LMEs of California Current and Gulf of California.
- U.S. East Coast (EC) – covering the LMEs of Northeast U.S. and Southeast U.S. continental shelf regions.
- Gulf of Mexico (GMx)

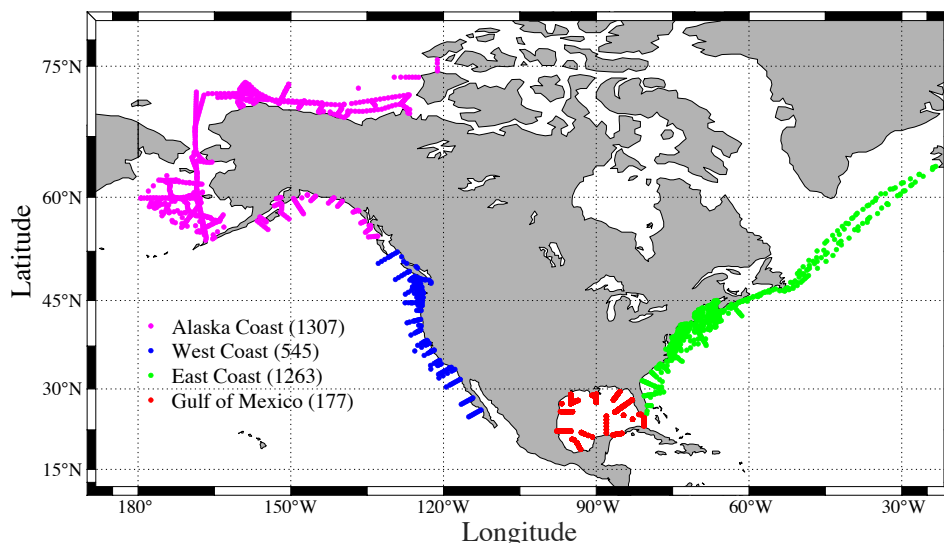
Data beyond continental shelves will be included if they are collected from a cruise that predominately covers parts of the U.S. North American ocean margins.

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Figure 1. A map showing all the sampling profiles of the CODAP-NA data product (v2020, a total of 3,292 profiles). Magenta dots show the sampling profiles in the Alaska Coast (AC). Blue ones are for the U.S. West Coast (WC), green ones are for the U.S. East Coast (EC), and the red ones are for the Gulf of Mexico (GMx). Numbers within the parentheses indicate the total number of profiles in the region.

3 Data sources

CODAP-NA was focused on chemical oceanographic data (inorganic carbon system parameters, oxygen, and nutrients) collected from discrete sampling-based observations. This also included discrete samples taken from shipboard flow-through systems rather than solely water collected in sampling rosette bottles. Carbon parameters recorded from continuous underway measurements by inline analytical instruments were excluded, as they had been QCed and included within the Surface Ocean CO_2 Atlas (SOCAT) (Bakker et al., 2016). The same was true for carbon parameters from time-series moorings. Data from large open estuaries (e.g., Salish Sea, Chesapeake Bay, Bay of Fundy) were also excluded during this first round of analysis, but these are among the data that may be able to benefit from secondary QC against CODAP.

We started with the highest quality coastal data sets to define a protocol for consistent QC and inter-comparison, which will subsequently be applied to other compiled coastal data sets. As a first step, only climate-quality discrete measurements (core data sets) with known quality and metadata from the Atlantic Oceanographic and Meteorological Laboratory (AOML), Pacific Marine Environmental Laboratory (PMEL), University of South Florida, University of Miami, University of Alaska Fairbanks, University of New Hampshire, and University of Delaware were included (Table 1). These data sets will serve as a reference for QCing future data sets.

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Table 1. List of cruises that are included in this version (v2020) of the CODAP-NA data product. Refer to Table 2 for the full names of the abbreviations and their units, and Table 3 for definitions of the Cruise_flags. CTD is short for conductivity, temperature, and depth, and refers to a package of electronic instruments that measure these properties. For samples collected from flow-through systems, temperature and salinity were also stored in CTDTEMP and CTDSAL, respectively. GMx is short for Gulf of Mexico.

NO	Region	EXPOCODE	Cruise_flag	CRUISE_ID	Start date	End date	Variables measured
1	Alaska Coast	33HQ20080329	B	HLY0802	2008-04-01	2008-05-06	CTDTEMP, CTDSAL, DIC, TALK, CTDOXY, Silicate, Phosphate, Nitrate, Nitrite, Ammonium
2	Alaska Coast	33HQ20080703	B	HLY0803	2008-07-04	2008-07-30	CTDTEMP, CTDSAL, DIC, TALK, CTDOXY, Silicate, Phosphate, Nitrate, Nitrite, Ammonium
3	Alaska Coast	33HQ20090403	B	HLY0902	2009-04-05	2009-05-10	CTDTEMP, CTDSAL, DIC, TALK, Silicate, Phosphate, Nitrate, Nitrite
4	Alaska Coast	33HQ20100907	B	HLY1003	2010-09-07	2010-09-08	CTDTEMP, CTDSAL, DIC, TALK
5	Alaska Coast	33HQ20111003	B	HLY1103	2011-10-06	2011-10-19	CTDTEMP, CTDSAL, TALK, pH
6	Alaska Coast	33HQ20121005	B	HLY1203	2012-10-10	2012-10-20	CTDTEMP, CTDSAL, DIC, TALK
7	Alaska Coast	33HQ20131005	D	HLY1303	2013-10-05	2013-10-30	CTDTEMP, CTDSAL, DIC, TALK
8	Alaska Coast	316N20090614	B	KN195	2009-06-22	2009-07-13	CTDTEMP, CTDSAL, DIC, TALK, Silicate, Phosphate, Nitrate
9	Alaska Coast	31FN20090924	B	MF0904	2009-09-26	2009-10-09	CTDTEMP, CTDSAL, DIC, TALK, CTDOXY, Silicate, Phosphate, Nitrate, Nitrite, Ammonium
10	Alaska Coast	33RO20150713	B	RB1504	2015-07-17	2015-07-31	CTDTEMP, CTDSAL, DIC, TALK, CTDOXY, Silicate, Phosphate, Nitrate, Nitrite
11	Alaska Coast	325020100509	B	TN249-10	2010-05-13	2010-07-12	CTDTEMP, CTDSAL, DIC, TALK, CTDOXY, Silicate, Phosphate, Nitrate, Nitrite, Ammonium
12	East Coast	316G20120202	C	DE1202	2012-02-06	2012-02-19	CTDTEMP, CTDSAL, DIC, TALK, pH, CTDOXY, Silicate, Phosphate, Nitrate_and_Nitrite, Ammonium
13	East Coast	33GG20150619	B	ECOA1	2015-06-20	2015-07-23	CTDTEMP, CTDSAL, DIC, TALK, pH, CTDOXY, Oxygen, Silicate, Phosphate, Nitrate, Nitrite, Nitrate_and_Nitrite
14	East Coast	33HH20180625	B	ECOA2	2018-06-25	2018-07-29	CTDTEMP, CTDSAL, DIC, TALK, pH, CTDOXY, Oxygen, Silicate, Phosphate, Nitrate_and_Nitrite
15	East Coast	334A20140510	C	EX1403	2014-05-10	2014-05-17	CTDTEMP, CTDSAL, DIC, TALK, pH, CTDOXY, Silicate, Phosphate, Nitrate_and_Nitrite, Ammonium
16	East Coast	33RO20070710	A	GOMECC1	2007-07-11	2007-08-02	CTDTEMP, CTDSAL, Salinity, DIC, TALK, CTDOXY, Oxygen, Silicate, Phosphate, Nitrate, Nitrite, Ammonium
17	East Coast	33RO20120721	A	GOMECC2	2012-07-22	2012-08-13	CTDTEMP, CTDSAL, Salinity, DIC, TALK, pH, Carbonate, CTDOXY, Oxygen, Silicate, Phosphate, Nitrate, Nitrite, Nitrate_and_Nitrite
18	East Coast	33GG20130609	C	GU1302	2013-06-09	2013-06-23	CTDTEMP, CTDSAL, DIC, TALK, pH, CTDOXY, Silicate, Phosphate, Nitrate_and_Nitrite, Ammonium
19	East Coast	33GG20131113	C	GU1305	2013-11-14	2013-11-24	CTDTEMP, CTDSAL, DIC, TALK, pH, CTDOXY, Silicate, Phosphate, Nitrate_and_Nitrite, Ammonium

20	East Coast	33GG20140301	C	GU1401 Leg2	2014-03-01	2014-03-08	CTDTEMP, CTDSAL, DIC, TALK, pH, Silicate, Phosphate, Nitrate_and_Nitrite, Ammonium
21	East Coast	33GG20151012	C	GU1506 Leg2	2015-10-13	2015-10-24	CTDTEMP, CTDSAL, DIC, TALK, pH, CTDOXY, Silicate, Phosphate, Nitrate_and_Nitrite, Ammonium
22	East Coast	33GG20160521	C	GU1608 Leg1	2016-05-23	2016-06-02	CTDTEMP, CTDSAL, DIC, TALK, pH, CTDOXY, Silicate, Phosphate, Nitrate_and_Nitrite, Ammonium
23	East Coast	33GG20160607	C	GU1608 Leg2	2016-06-08	2016-06-12	CTDTEMP, CTDSAL, DIC, TALK, pH, CTDOXY, Silicate, Phosphate, Nitrate_and_Nitrite, Ammonium
24	East Coast	33GG20170516	C	GU1701 Leg1	2017-05-17	2017-05-25	CTDTEMP, CTDSAL, DIC, TALK, pH, CTDOXY, Silicate, Phosphate, Nitrate_and_Nitrite, Ammonium
25	East Coast	33GG20170530	C	GU1701 Leg2	2017-05-31	2017-06-05	CTDTEMP, CTDSAL, DIC, TALK, pH, CTDOXY, Silicate, Phosphate, Nitrate_and_Nitrite, Ammonium
26	East Coast	33GG20170610	C	GU1702	2017-06-12	2017-06-21	CTDTEMP, CTDSAL, DIC, TALK, pH
27	East Coast	33GG20171031	C	GU1706	2017-11-01	2017-11-07	CTDTEMP, CTDSAL, DIC, TALK, pH, CTDOXY, Silicate, Phosphate, Nitrate_and_Nitrite, Ammonium
28	East Coast	33GG20180822	C	GU1804	2018-08-23	2018-08-29	CTDTEMP, CTDSAL, DIC, TALK, pH, CTDOXY, Silicate, Phosphate, Nitrate_and_Nitrite
29	East Coast	33HH20120531	C	HB1202	2012-06-02	2012-06-13	CTDTEMP, CTDSAL, DIC, TALK, pH, CTDOXY, Silicate, Phosphate, Nitrate_and_Nitrite, Ammonium
30	East Coast	33HH20130314	D	HB1301	2013-03-17	2013-05-09	CTDTEMP, CTDSAL, DIC, TALK, pH, Silicate, Phosphate, Nitrate_and_Nitrite, Ammonium
31	East Coast	33HH20140908	D	HB1405 Leg1	2014-09-10	2014-09-18	CTDTEMP, CTDSAL, DIC, TALK, pH
32	East Coast	33HH20140923	D	HB1405 Leg2	2014-09-25	2014-09-30	CTDTEMP, CTDSAL, DIC, TALK, pH
33	East Coast	33HH20141028	D	HB1405 Leg4	2014-11-04	2014-11-05	CTDTEMP, CTDSAL, DIC, TALK, pH
34	East Coast	33HH20150519	C	HB1502	2015-05-20	2015-06-02	CTDTEMP, CTDSAL, DIC, TALK, CTDOXY, Silicate, Phosphate, Nitrate_and_Nitrite, Ammonium
35	East Coast	33HH20170211	C	HB1701	2017-02-12	2017-02-22	CTDTEMP, CTDSAL, DIC, TALK, pH, CTDOXY, Silicate, Phosphate, Nitrate_and_Nitrite, Ammonium
36	East Coast	33HH20180523	C	HB1803	2018-05-23	2018-06-04	CTDTEMP, CTDSAL, DIC, TALK, pH, CTDOXY, Silicate, Phosphate, Nitrate_and_Nitrite, Ammonium
37	East Coast	334B20121026	C	PC1207	2012-10-27	2012-11-13	CTDTEMP, CTDSAL, DIC, TALK, CTDOXY, Silicate, Phosphate, Nitrate_and_Nitrite, Ammonium
38	East Coast	334B20141103	C	PC1405	2014-11-04	2014-11-18	CTDTEMP, CTDSAL, DIC, TALK, pH, CTDOXY, Silicate, Phosphate, Nitrate_and_Nitrite, Ammonium
39	East Coast	334B20160807	C	PC1604	2016-08-09	2016-08-19	CTDTEMP, CTDSAL, DIC, TALK, pH, CTDOXY, Silicate, Phosphate, Nitrate_and_Nitrite, Ammonium
40	East Coast	334B20161018	C	PC1609	2016-10-19	2016-10-19	CTDTEMP, CTDSAL, DIC, TALK, pH, CTDOXY, Silicate, Phosphate, Nitrate_and_Nitrite, Ammonium
41	East Coast	33H520181102	C	S11802	2018-11-02	2018-11-12	CTDTEMP, CTDSAL, DIC, TALK, pH, CTDOXY, Silicate, Phosphate, Nitrate_and_Nitrite, Ammonium
42	East Coast	AGSK20031205	D	SKO0313	2003-12-06	2003-12-14	CTDTEMP, CTDSAL, DIC
43	East Coast	AGSK20040403	D	SKO0406	2004-04-04	2004-04-11	CTDTEMP, CTDSAL, DIC
44	East Coast	AGSK20040625	D	SKO0410	2004-06-26	2004-07-02	CTDTEMP, CTDSAL, DIC, TALK
45	East Coast	AGSK20041015	D	SKO0414	2004-10-16	2004-10-22	CTDTEMP, CTDSAL, DIC, TALK
46	East Coast	AGSK20050916	D	SKO0510	2005-09-17	2005-09-23	CTDTEMP, CTDSAL, DIC, TALK
47	East Coast	AGSK20060403	D	SKO0604	2006-04-18	2006-04-27	CTDTEMP, CTDSAL, DIC, TALK
48	East Coast	AGSK20061014	D	SKO0611	2006-10-15	2006-10-23	CTDTEMP, CTDSAL, DIC, TALK

49	East Coast	AGSK20070525	D	SKO0721	2007-05-26	2007-06-02	CTDTEMP, CTDSAL, DIC
50	East Coast	AGFO20140607	D	SKO1406	2014-06-20	2014-06-28	CTDTEMP, CTDSAL, DIC, TALK, pH
51	East Coast	46SL20181115	D	Selfoss846	2018-11-15	2018-11-22	CTDTEMP, CTDSAL, DIC, TALK, pH, Silicate, Phosphate, Nitrate_and_Nitrite
52	GMx	33RO20170718	A	GOMECC3	2017-07-18	2017-08-20	CTDTEMP, CTDSAL, Salinity, DIC, TALK, pH, Carbonate, CTDOXY, Oxygen, Silicate, Phosphate, Nitrate, Nitrite, Nitrate_and_Nitrite
53	GMx	33WA20141201	C	WS1418	2014-12-03	2014-12-04	CTDTEMP, CTDSAL, DIC, TALK, pH, CTDOXY, Silicate, Phosphate, Nitrate_and_Nitrite
54	GMx	33WA20150921	C	WS15264	2015-09-23	2015-09-24	CTDTEMP, CTDSAL, DIC, TALK, pH, Silicate, Phosphate, Nitrate_and_Nitrite
55	West Coast	332220170918	A	SH1709	2017-09-18	2017-09-28	CTDTEMP, CTDSAL, DIC, TALK, CTDOXY, Oxygen, Silicate, Phosphate, Nitrate, Nitrite, Ammonium
56	West Coast	32WC20070511	A	WCOA2007	2007-05-14	2007-06-12	CTDTEMP, CTDSAL, Salinity, DIC, TALK, CTDOXY, Oxygen, Silicate, Phosphate, Nitrate, Nitrite
57	West Coast	32WC20110812	A	WCOA2011	2011-08-12	2011-08-30	CTDTEMP, CTDSAL, Salinity, DIC, TALK, pH, CTDOXY, Oxygen, Silicate, Phosphate, Nitrate, Nitrite, Ammonium
58	West Coast	332220120904	A	WCOA2012	2012-09-05	2012-09-16	CTDTEMP, CTDSAL, Salinity, DIC, TALK, CTDOXY, Oxygen, Silicate, Phosphate, Nitrate, Nitrite, Ammonium
59	West Coast	317W20130803	A	WCOA2013	2013-08-05	2013-08-10	CTDTEMP, CTDSAL, Salinity, DIC, TALK, pH, Carbonate, CTDOXY, Oxygen, Silicate, Phosphate, Nitrate, Nitrite, Ammonium
60	West Coast	32P020130821	A	WCOA2013	2013-08-21	2013-08-28	CTDTEMP, CTDSAL, Salinity, DIC, TALK, pH, Carbonate, CTDOXY, Oxygen, Silicate, Phosphate, Nitrate, Nitrite, Ammonium
61	West Coast	33RO20160505	A	WCOA2016	2016-05-08	2016-06-06	CTDTEMP, CTDSAL, Salinity, DIC, TALK, pH, Carbonate, CTDOXY, Oxygen, Silicate, Phosphate, Nitrate, Nitrite, Ammonium

4 Parameters / variables

For the current version of the CODAP-NA, inorganic carbon system parameters, oxygen, nutrients, and related hydrographic parameters were included (Table 2). CTDPRES, CTDTEMP, CTDSAL, and CTDOXY were commonly measured with pressure, temperature, conductivity, and oxygen sensors, respectively, mounted on a CTD rosette. In some cruises with

160 surface samples collected from flow-through systems, temperature and salinity were also provided in columns reserved for CTDTEMP and CTDSAL, respectively. Water samples were collected and measured onboard or later in a shore-based laboratory for discrete salinity, discrete dissolved oxygen concentration (DO), dissolved inorganic carbon concentration (DIC), total alkalinity (TALK), pH, carbonate ion concentration ($[CO_3^{2-}]$), fugacity of carbon dioxide (fCO_2), and concentrations of silicic acid, phosphate, nitrate, nitrite, nitrate plus nitrite, and ammonium. For discrete pH on the Total

165 Scale, $[CO_3^{2-}]$, and fCO_2, both measured and calculated values were presented. Saturation states of aragonite (Ω_{arag}) and calcite (Ω_{calc}) could only be calculated. The carbonate system calculations were conducted using the MATLAB version 3.01 (Sharp et al., 2020) of the CO2SYS program (Lewis and Wallace, 1998), with the dissociation constants for carbonic acid of

Lueker et al. (2000), bisulfate (HSO₄⁻) of Dickson (1990), hydrofluoric acid (HF) of Perez and Fraga (1987), and with the

total borate equations of Lee et al., (2010).

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Table 2. Parameters that are included in the CODAP-NA (v2020) data product.

Abbreviation	Variable name	Unit	Measured/ calculated
CTDPRES	Water pressure recorded from sensors on a CTD rosette. For surface samples collected from an onboard flow-through system, its pressure is equal to the depth of the water inlet. When such info is not available, it is assumed to be 5 dbar.	dbar	measured
CTDTEMP_ITS90	Temperature on the International Temperature Scale of 1990 (ITS-90) from sensors on a CTD rosette. For surface samples collected from an onboard flow-through system, temperature has also been merged into the CTDTEMP_ITS90 variable.	°C	measured
CTDSAL_PSS78	Salinity on the Practical Salinity Scale 1978 (PSS-78) from sensors on a CTD rosette. For surface samples collected from an onboard flow-through system, salinity from the thermosalinograph (TSG) has been merged into the CTDSAL_PSS78 variable.	-	measured
Salinity_PSS78	Salinity on the PSS-78 scale measured as discrete samples	-	measured
recommended_Salinity_PSS78	Discrete salinity with some missing values filled in using CTDSAL	-	measured
CTDOXY	Dissolved oxygen concentration from sensors on a CTD rosette	µmol kg⁻¹	measured
Oxygen	Dissolved oxygen concentration measured as discrete samples from Winkler titration	µmol kg⁻¹	measured
recommended_Oxygen	Discrete dissolved oxygen concentration from Winkler titration with some missing values filled in using CTDOXY	µmol kg⁻¹	measured
AOU	Apparent oxygen utilization	µmol kg⁻¹	calculated
DIC	Dissolved inorganic carbon concentration	µmol kg⁻¹	measured
TALK	Total alkalinity	µmol kg⁻¹	measured
pH_TS_measured	pH on total hydrogen scale (TS) measured as discrete samples at measurement temperature and ambient pressure	-	measured
TEMP_pH	Temperature of pH measurement	°C	measured
pH_TS_insitu_measured	pH on total hydrogen scale (TS) measured as discrete samples and adjusted to *in-situ* conditions	-	measured
pH_TS_insitu_calculated	pH on total hydrogen scale (TS) at *in-situ* conditions calculated from DIC, TA and other parameters using CO2SYS	-	calculated
Carbonate_measured	Carbonate ion concentration measured as discrete samples at measurement temperature and ambient pressure	µmol kg⁻¹	measured
TEMP_Carbonate	Temperature of carbonate ion measurement	°C	measured
Carbonate_insitu_measured	Carbonate ion concentration measured as discrete samples and adjusted to *in-situ* conditions	µmol kg⁻¹	measured
Carbonate_insitu_calculated	Carbonate ion concentration at *in-situ* conditions calculated from DIC, TA and other parameters using CO2SYS	µmol kg⁻¹	calculated
fCO₂_measured	Fugacity of carbon dioxide measured as discrete samples at measurement temperature and ambient pressure	µatm	measured
TEMP_fCO₂	Temperature of fCO₂ measurement	°C	measured
fCO₂_insitu_measured	Fugacity of carbon dioxide measured as discrete samples and adjusted to *in-situ* conditions	µatm	measured
fCO₂_insitu_calculated	Discrete fugacity of carbon dioxide at *in-situ* conditions calculated from DIC, TA and other parameters using CO2SYS	µatm	calculated
Aragonite	Aragonite saturation state at *in-situ* conditions calculated from DIC, TA and other parameters using CO2SYS	-	calculated

Calcite	Calcite saturation state at *in-situ* conditions calculated from DIC, TA and other parameters using CO2SYS	-	calculated
Revelle_Factor	Revelle Factor calculated from DIC, TA and other parameters using CO2SYS	-	calculated
Silicate	Silicate concentration	µmol kg^{-1}	measured
Phosphate	Phosphate concentration	µmol kg^{-1}	measured
Nitrate	Nitrate concentration	µmol kg^{-1}	measured
Nitrite	Nitrite concentration	µmol kg^{-1}	measured
Nitrate_and_Nitrite	Nitrate and Nitrite concentrations combined	µmol kg^{-1}	measured
recommended_Nitrate_and_Nitrite	Nitrate_and_Nitrite concentration, along with Nitrate concentration when Nitrate_and_Nitrite data are not available	µmol kg^{-1}	measured
Ammonium	Ammonium concentration	µmol kg^{-1}	measured

5 Technical Approach and Methodology

Quality control often involves two steps: primary QC and secondary QC (Tanhua et al., 2010). Primary QC is the process of identifying outliers and obvious errors within an individual cruise data set using measurement metadata or approaches like
175 property-to-property plots. It should largely be done by the investigators responsible for the measurements. However, it is advisable to provide additional uniform primary QC to all cruises within a data product using common tools and common thresholds to help identify any issues that have been missed by the data producers. These issues are communicated back to the investigators so that the issues could be reviewed and, if necessary, addressed. This additional layer of primary QC is often performed by the data product synthesis community. Secondary QC is a process in which data from one cruise are
180 objectively compared against data from another cruise or a previously synthesized dataset in order to quantify systematic differences in the reported values. The secondary QC process often entails cross-over analysis (Lauvset and Tanhua, 2015), and increasingly regional Multiple Linear Regression (MLR) and inversions (Olsen et al., 2019; 2020).

Due to the scarcity of cross-over stations at depths where parameters were not influenced by temporal variations (sampling
185 depth >1500 m, Olsen et al., 2020) on coastal cruises, secondary QC was not conducted for this version of the CODAP-NA and no cruise-wide offsets or multiplicative adjustments were applied. Instead, the QC relied on (a) stringent criteria for the selection of data sources, and (b) an enhanced primary QC procedure with rigorous consistency checks. This version of the CODAP-NA only accepted data from laboratories with direct involvement in the CODAP effort and with a track record of producing high-quality data and following best practices, making secondary quality control less essential. It is likely that
190 there are other very high-quality coastal cruise data sets that are not yet included in this version of CODAP-NA.

We worked directly with the data providers who knew their data best to conduct these primary QC procedures in order to leverage all of the resources related to a measurement: details related to the methods, instrumentation, reference standards, access to the raw data, and the analysts' recollection of the measurements. A new suite of QC tools was developed by this
195 team of authors to satisfy the requirements of enhanced consistency checks. These tools will be made available to the public

soon, with a separate paper dedicated to their rationales, development details, and instructions (Jiang et al., in prep.). The plan is to make it available through a web interface, so that no MATLAB license is required to use the tool. Below are the major steps of the QC procedures:

200 **Step One** was to ensure all of the cruise data files were ingested into NCEI's archives and documented with a rich metadata record (Jiang et al., 2015b). Maintaining a cruise data table allowing future users of the data product to access the original data files is an important component of any synthesis effort. For this study, a table with key metadata is available through this link: https://www.ncei.noaa.gov/access/ocean-acidification-data-stewardship-oads/synthesis/NAcruises.html. The following fields are listed in the table: A sequential number of the individual cruise data set (NO), expedition code

205 (EXPOCODE), flags indicating the quality of the cruise (Cruise_flag, see Table 3), cruise identifier (Cruise_ID), Start_date, End_date, measured parameters, and links to NCEI's archive) .

Step Two was to load the measurement values from the original cruise data files into MATLAB. All missing values were replaced with "-999" during this process. Variables without a QC flag from the original cruise data file were assigned an

210 initial flag of 2 (good values, Table 4). Variables that were clearly out of range (e.g., a DIC value of < 0) were automatically assigned with a QC flag of "4" (bad values). The QC flags for all "-999" values or missing values were replaced with "9" (missing values).

Some surface samples from a few coastal cruises were collected from flow-through systems onboard research vessels,

215 instead of Niskin bottles on sampling rosettes. In such cases, the temperature and salinity values were stored under the CTDTEMP and CTDSAL columns, respectively, although they were not measured from sensors mounted on a CTD rosette. Similarly, their sampling depth values were extracted from the metadata as the depth of the water inlet and stored under CTDPRES (Table 2). When water inlet depth information was not available, its sampling depth was set to be 5 dbar. There is a column named "Observation_type" in the CODAP data product file to indicate whether a sample is from a "Flow-through"

220 system or a "Niskin" bottle.

Step Three was to conduct several key calculations. The QC tool automatically calculated or assigned the below parameters:
    (a) Sample_ID if not already included (Equation 1)
    (b) depth from pressure and vice versa;
225   (c) recommended_Salinity_PSS78 (Table 2);
    (d) conservative temperature, absolute salinity, sigma-theta;
    (e) recommended_Oxygen
    (f) apparent oxygen utilization (AOU);
    (g) recommended_Nitrate_and_Nitrite;

230   (h)   calculated pH, carbonate ion, and fCO$_2$ at *in-situ* conditions using CO2SYS from DIC and TALK, along with
              temperature, salinity, pressure, and nutrients; and

        (i)   *in-situ* pH, carbonate ion, and fCO$_2$ from their respective values at their measurement conditions.

        Sample_IDs were calculated from STATION_ID (station identification number), CAST_NO (cast number) and NISKIN_ID
235   (Niskin identification) based on equation (1), if they were not already available:

              Sample_ID = Station_ID × 10000 + Cast_number × 100 + Niskin_ID   (1)

        For example, at station 15, the 2nd cast, a Niskin_ID of 3 will have a Sample_ID of 150203. In cases when they could not be
240   calculated (e.g., Station_ID is non-numerical), Sample_ID was assigned as 1, 2, 3, … from the first row to the last row of the
        original cruise data file.

        Sampling depth (Depth) and pressure (CTDPRES) were calculated from one another where applicable using the equations of
        "gsw_z_from_p", and "gsw_p_from_z", respectively, from the International Thermodynamic Equation of Seawater 2010
245   (TEOS-10; IOC et al., 2010). When both values were available, CTDPRES values were preferentially used, and the
        calculated Depth values were used to replace the original Depth values.

        The "recommended_salinity_PSS78" column was created by merging the discrete salinity and CTDSAL columns. Data were
        preferentially chosen from the discrete measurements provided their QC flags were equal to 2 or 6. If these values were not
250   available, CTDSAL values with QC flags of 2 or 6 were chosen. In the absence of these two, discrete salinity measures with
        QC flags other than 2 or 6 were chosen. Lastly, the CTDSAL values with other QC flags were chosen. The same principles
        were applied to merge the oxygen data. The merged discrete oxygen and CTDOXY data were stored in the column named
        "recommended_Oxygen. (Table 2).

255   Conservative temperature (Θ) is proportional to the potential enthalpy and is recommended as a replacement for potential
        temperature (θ), as it more accurately represents the heat content (IOC et al., 2010). Absolute Salinity (S_A) is the mass
        fraction of salt in seawater (unit: g/kg) based on conductivity ratio plus a regional correction term as opposed to the practical
        salinity scale (SP, Practical Salinity Scale 1978, or PSS-78, unitless, based solely on the conductivity ratio) (Le Menn et al.,
        2018). Conservative temperature, absolute salinity, and sigma-theta were calculated using the equations of
260   "gsw_CT_from_t", "gsw_SA_from_SP", and "gsw_sigma0", respectively, from the TEOS-10 (IOC et al., 2010). Apparent
        oxygen utilization (AOU) was calculated based on absolute salinity, conservative temperature, latitude, longitude,
        CTDPRES, and recommended_Oxygen variable using the function "gsw_O2sol" as described in the TEOS-10 (IOC et al.,
        2010). Oxygen solubility is estimated with the combined equation from Garcia and Gordon (1992).

265   In order to measure nitrate, it is first reduced to nitrite and then nitrate plus nitrite is measured together (Hydes and Hill, 1985). The concentration of nitrite is usually much lower than nitrate. When nitrite is not reported, it is often because its concentration is too low to be detectable. For the CODAP-NA data product, when Nitrate values were not available, but both Nitrate_and_Nitrite and Nitrite values with QC flags of 2 or 6 were available, Nitrate values were calculated by subtracting Nitrite from Nitrate_and_Nitrite. Similarly, when Nitrate_and_Nitrite values were not available, but both Nitrate and Nitrite

270   values with QC flags of 2 or 6 were available, Nitrate_and_Nitrite values were calculated by adding Nitrate and Nitrite concentrations together. The "recommended_Nitrate_and_Nitrite" column was created by preferentially using Nitrate_and_Nitrite values. In cases when Nitrate_and_Nitrite values were not available but Nitrate values with a QC flag of 2 or 6 were available (Nitrite values not available), the Nitrate_and_Nitrite values were assumed to be approximate to the Nitrate values.

275

Carbonate_insitu_measured, pH_TS_insitu_measured, and fCO$_2$_insitu_measured (Table 2) were recalculated from their respective values at measurement conditions (i.e., pH_TS_measured, Carbonate_measured, and fCO$_2$_insitu_measured) with the CO2SYS program, using the dissociation constants as described above. TALK was preferentially used as the second carbon parameter. When it was not available, DIC was used. If neither of them was available, TALK derived from salinity

280   with the locally interpolated alkalinity regression (LIARv2) method was used for the adjustment from measurement to *in-situ* conditions (Carter et al., 2018). Carbonate_insitu_calculated, pH_TS_insitu_calculated, fCO$_2$_insitu_calculated, aragonite saturation state, calcite saturation state, and Revelle_Factor were calculated from DIC and TALK, along with *in-situ* temperature, salinity, pressure, silicate, and phosphate using the same dissociation constants as above (Table 2). When either silicate or phosphate data were unavailable, their mean values during the cruise were used for the calculation. Samples

285   with a salinity of less than 15 were excluded from this calculation, due to the potentially large uncertainties.

Step Four was to identify outliers. Outliers were determined by visual inspection. Two types of outlier identification were used for this effort: (a) a broad-scale outlier identification by visually examining the plot of a variable against its sampling depth and other property-to-property plots, and (b) a fine-scale outlier identification based on consistency checks. Here,

290   consistency checks refer to both the "internal consistency checks", i.e., the comparison of a measurement with its calculated value (e.g., spectrophotometrically-measured pH vs. pH calculated from other carbon parameters using CO2SYS), as well as a measurement with one method against that with a different method (e.g., oxygen measured from Winkler vs. a sensor). For the broad-scale outlier identification, the QC tools make plots of all variables against depth (or sigma-theta when only surface values are available), as well as these plots:

295   (a)   TALK against salinity,
            (b)   DIC against dissolved oxygen (DO),
            (c)   all nutrients (silicate, phosphate, nitrate, nitrite, nitrate plus nitrite, ammonium) against DO.

Consistency check-based outlier identification was the primary way of finding outliers in this study. Consistency checks
were conducted for these variable pairs:

(a) CTDSAL vs. discrete salinity

(b) CTDOXY vs. discrete oxygen measured from Winkler titration

(c) pH measured with a spectrophotometer vs. pH calculated with CO2SYS from DIC, TALK and other parameters

(d) Carbonate ion ($[CO_3^{2-}]$) measured with a spectrophotometer vs. $[CO_3^{2-}]$ calculated with CO2SYS from DIC, TALK and other parameters

(e) Discrete fCO_2 measured with a non-dispersive infrared analyzer vs. fCO_2 calculated with CO2SYS from DIC, TALK and other parameters.

Step five was to append all of the individual cruise data files one after another into one data product file with all of the variables as listed in Table 2. For surface samples collected from flow-through systems, their Cast_numbers and Niskin_IDs were all set to "-999", and their Niskin_flags were all set to "9". The contents of Observation_type were standardized to be either "Niskin" or "Flow-through". Data values with QC flags that were not 2 (good), 3 (questionable), or 6 (average of duplicate measurements) were replaced with "-999", and their corresponding QC flags were changed to "9". The merged data product file was further QCed by plotting all of the non-missing values for each variable. These plots were examined further, with focus on the outliers falling out of 2.5 times their respective standard deviations.

6 Data products

The data product is available in Excel, CSV, MATLAB, and NetCDF formats at NOAA/NCEI with a DOI of [10.25921/531n-c230] and NCEI Accession Number of [0219960] (Jiang et al., 2020). All parameters in Table 2, along with their primary level QC flags (Table 4) and Cruise_flags (Table 3) are presented. The chosen primary level QC flag convention is the same as the GLODAPv2 project (Olsen et al., 2020). Note the difference between the WOCE primary level QC flags (e.g., 2, 3, 4, 9, etc.) and the Secondary QC flags as used by the GLODAPv2 (a choice of either 0 or 1). The "cruise flags" were newly minted to indicate the overall quality of a cruise data set (Table 3). In the current version (v2020) of the CODAP-NA, there are 3,292 discrete chemical oceanographic profiles, and a total of 27,404 data points. They were collected on 61 cruises in the ocean margins of North America from December 6, 2003 to November 22, 2018. There are on average eight sampling depth levels (a median of seven) for each profile. The total count of data points for each parameter and their minimum, maximum, and mean values are listed in Table 5.

Table 3. Cruise flags used for this product.

Flag value	Meaning
A	These were dedicated OA cruises that were executed following Best Practices for global ocean work as outlined in Hood et al. (2011) and other documents as can be found on GO-SHIP site[*]. Colloquially these are referred to as GO-SHIP quality. Traceable standards and certified reference materials were used, and deep stations (> 2500 m) were sampled to allow using near-constant deep-water concentrations as anchor points. A third inorganic carbon system parameter, such as pH or carbonate ion concentration were often measured, allowing consistency checks.
B	These are dedicated OA cruises that had onboard inorganic carbon measurements performed according to Best Practices (Dickson et al. 2007), and many other parameters to highest accuracy through use of standards and certified reference materials. However, the cruises did not necessarily have all other parameters analyzed to highest standards, such as freezing nutrients for shoreside analyses; not taking oxygen and nutrients samples on most Niskins; not normalizing CTD/O_2 trace to Winkler oxygen values, insufficient metadata etc. There often are insufficient deep stations to compare data with open ocean data.
C	These were opportunistic cruises where OA parameters were measured in the water column. They include standard hydrographic, carbon, and OA parameters; T, S, O_2, nutrients, TALK, DIC, pH. Many parameters, including carbon and OA parameters were measured shoreside; CTD oxygen data were not adjusted to Winkler oxygen values. Generally, no dedicated OA personnel were onboard.
D	Underway samples only. These cruises had no CTD casts, and only had samples taken from the seawater supply line, with often a limited amount of other hydrographic parameters. T and S were obtained from thermosalinographs with limited or no salinity check samples.

([*]https://www.go-ship.org/HydroMan.html)

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Table 4. World Ocean Circulation Experiment (WOCE) World Hydrographic Program (WHP) (Joyce and Corry, 1994; Swift and Diggs, 2008) QC flags used for this product.

Flag value	Meaning
2	Acceptable
3	Questionable
6	Average of duplicates
9	Missing value

335 **Table 5.** The minimum, maximum, mean, and data point counts of the parameters that are included in the final product. Refer to Table 2 for their full parameter names and units.

Abbreviation	Min	Max	Mean	Count
CTDTMP_ITS90	-1.79	31.74	10.62	27,382
CTDSAL_PSS78	1.38	37.61	33.36	27,263
Salinity_PSS78	0.12	36.97	34.8	3,488
recommended_Salinity_PSS78	0.12	37.61	33.36	27,277
CTDOXY	3.1	481.7	216.7	21,750
Oxygen	2.7	472	183.7	11,780
recommended_Oxygen	2.7	481.7	216.8	22,335
AOU	-190.6	313.5	62.6	22,200
DIC	886.4	2621	2103	19,087
TALK	878	2853	2249	19,088
pH_TS_insitu_measured	7.41	8.58	7.94	9,897
pH_TS_insitu_calculated	7.42	8.57	7.93	17,714
Carbonate_insitu_measured	33.3	298.9	129.4	4,621
Carbonate_insitu_calculated	21.4	306.9	113.3	17,714
fCO2_insitu_measured	198.3	1175	450.8	3,140

fCO2_insitu_calculated	90.6	1788	568.4	17,714
Aragonite	0.34	5.1	1.71	17,714
Calcite	0.58	7.67	2.66	17,714
Revelle_Factor	8.26	19.7	14.35	17,714
Silicate	0	234.4	24.3	19,480
Phosphate	0	3.64	1.22	19,452
Nitrate	0	51.06	15.36	16,508
Nitrite	0	3.76	0.09	15,376
Nitrate_and_Nitrite	0	51.06	14.05	18,069
recommended_Nitrate_and_Nitrite	0	51.06	14.08	19,330
Ammonium	0	7.92	0.58	10,940

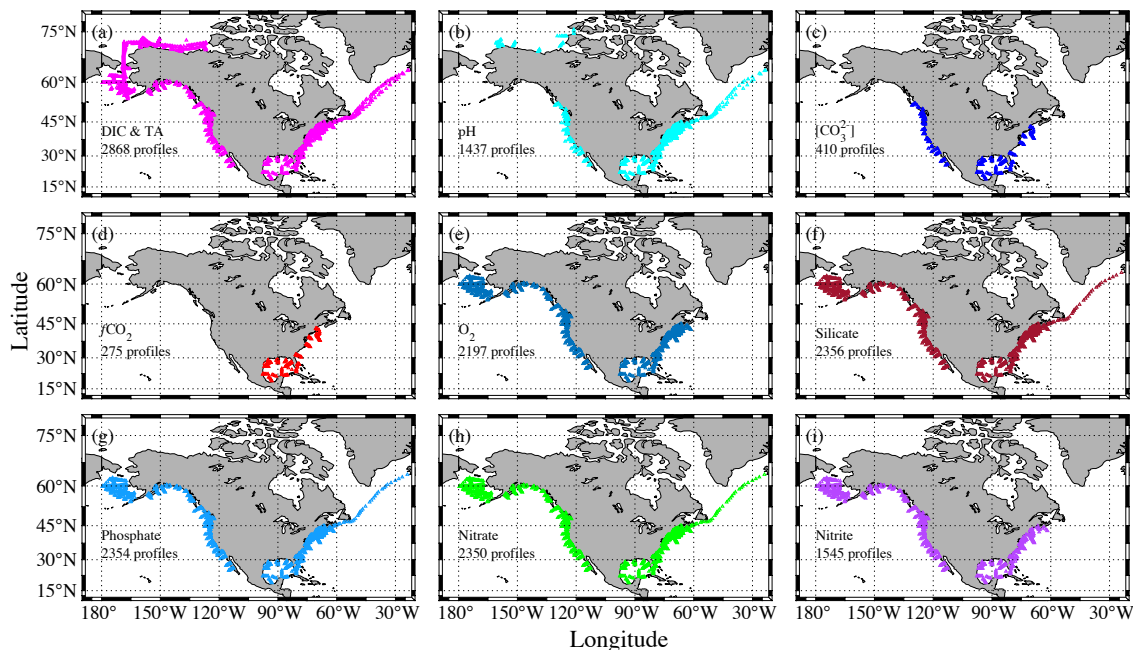
Of the 3,292 profiles, 2,868 have both DIC and TALK measurements, thus the full list of carbonate system parameters (pH, fCO$_2$, [CO$_3^{2-}$], aragonite saturation state, calcite saturation state, and Revelle Factor) can be calculated (Figure 2). In addition, there are 1,437 profiles with discrete pH measurements from a spectrophotometer-based method (Byrne and Breland, 1989; Clayton and Byrne, 1993; Dickson, 1993), 410 profiles with discrete carbonate ion measurements (Byrne and Yao, 2008; Sharp and Byrne, 2019), and 275 profiles with discrete fCO$_2$ measurements (Wanninkhof and Thoning, 1993). There is also good coverage of oxygen and nutrients measurements (Figure 2).

345

Figure 2. Sampling profiles for certain parameters. A profile is plotted if it has at least one measured value. Panel (a) only includes profiles that have both dissolved inorganic carbon (DIC) and total alkalinity (TA) measured. Panel (b) is for profiles with discrete pH measurements from a spectrophotometer. Panel (c) is for profiles with discrete carbonate ion concentration ([CO$_3^{2-}$]) measurements from a spectrophotometer. Panel (d) is for profiles with discrete

350 **fugacity of carbon dioxide (*f*CO₂)measurements from flasks. Panel (e) is for profiles with either recommended_Oxygen values (see Table 2 for more details). Panels (f-i) are for profiles with nutrient measurements.**

One major difference between the CODAP-NA and the GLODAPv2 is the shallower sampling depths of the former (Figure 3). About 80% of the 3,292 profiles have a maximum sampling depth of < 300 m, and 30% of them have maximum

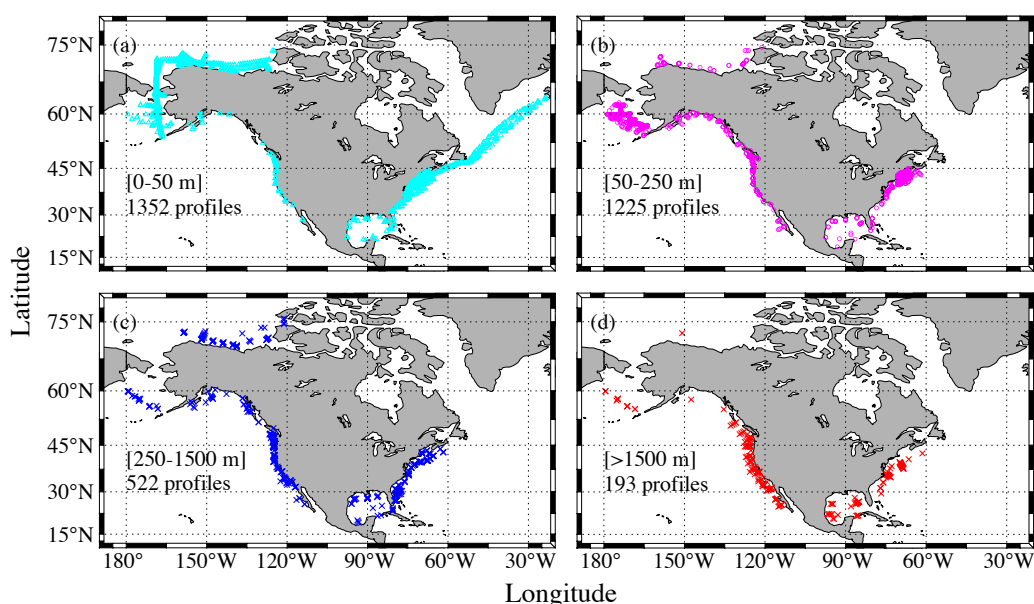
355 sampling depth of < 25m, with a lot of them being surface-only measurements. Only 193 profiles (< 6% of the total 3,292 profiles) have at least one sampling depth level below 1500 m, which has commonly been used as a threshold for subsurface cross-over analyses (Figure 3). Most of these deep-water profiles are found off the U.S. West Coast, Gulf of Mexico, and a few offshore stations in the Mid-Atlantic Bight. On average, the sampling depth is 300 m, with a median sampling depth of only 65 m.

360

Figure 3. Sampling depths of the profiles: (a) profiles with maximum depths ranging from 0 to 50 m, (b) profiles with maximum depths ranging from 50 to 250 m, (c) profiles with maximum depths ranging from 250 to 1500 m, (d) profiles with maximum depths greater than 1500 m.

365 Another distinctive feature of coastal oceans is their large magnitude of seasonal variation. For a lot of parameters, their seasonal variation, along with the diel and intertidal variations often eclipse their long-term variation. Understanding the seasonal variation and de-seasonalizing the observation data are often critical steps in the process of deciphering the long-term change. Like most data products, this version of the CODAP-NA is summer- and fall-biased, with spring, summer, fall

and winter having 677, 1538, 975, and 102 profiles, respectively (Figure 4). All coasts have good summer data coverage, but
370   the only area with meaningful winter data coverage is the northeastern U.S. coast (Figure 4, Table 6).

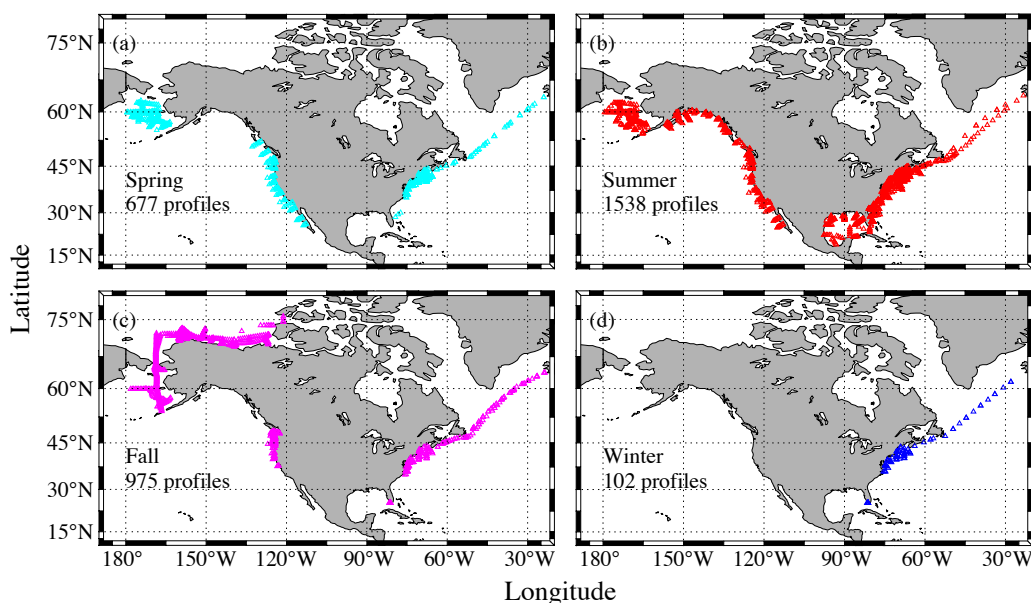



Figure 4. Sampling profiles in each of the four seasons: (a) Spring (March – May), (b) Summer (June – August), (c) Fall (September – November), (d) Winter (December – February).

Table 6. Number of profiles and data points (the sum of all depth levels at each profile) in all seasons of each region.

	Spring (Mar – May)		Summer (Jun – Aug)		Fall (Sep – Nov)		Winter (Dec – Feb)	
	Profiles	Data points	Profiles	Data points	Profiles	Data points	Profiles	Data points
Alaska Coast	301	2053	409	2762	597	1818	0	0
West Coast	167	3024	250	3874	128	1524	0	0
East Coast	209	484	728	8781	235	554	91	235
Gulf of Mexico	0	0	151	2269	15	15	11	11

375

To demonstrate the large seasonal amplitude (defined here as the difference between the maximum and minimum values of a variable on an annual cycle) in the study area, an analysis was conducted to group surface stations (with at least one sampling depth < 25 m) that are within 1 km distance and have at least one measurement between December and March and one measurement between June and October. The results, which are based on 135 groups of stations (most of them in the
380   northeastern U.S. coast), show large seasonal variations for nearly all the variables (Figure 5). The average seasonal amplitudes, and their percentage changes are: CTDTEMP (13.9 °C), CTDSAL (2.3, 7%), TALK (112 µmol kg^{-1}, 5%), DIC (126 µmol kg^{-1}, 6%), fCO$_2$ (170 µatm, 39%), [CO$_3^{2-}$] (61 µmol kg^{-1}, 45%), pH (0.16, 2%), aragonite saturation state (0.99, 47%), calcite saturation state (1.47, 45%). Note the "seasonal amplitudes" here represent the sum of effects of all changes

including changes from freshwater input, mixing, upwelling, warming and cooling, biological cycling, and diurnal cycling
385 within a season.

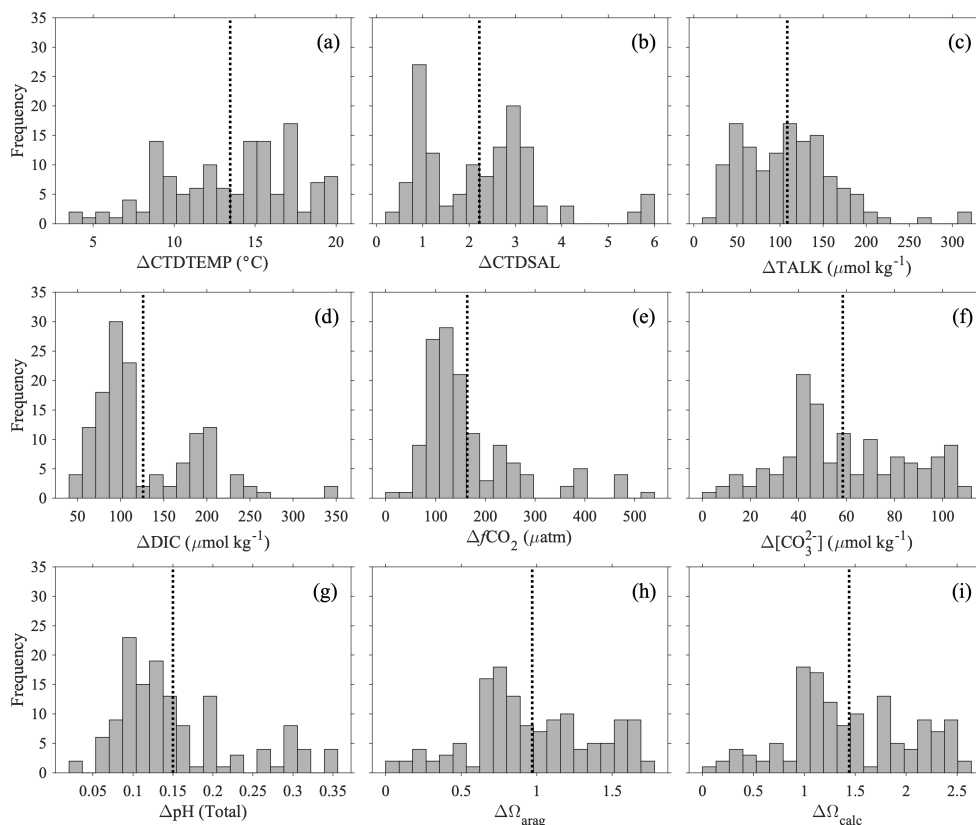



Figure 5. Seasonal amplitudes (maximum minus minimum values within a group of close by stations) of (a)
temperature (CTDTEMP), (b) salinity (CTDSAL), (c) total alkalinity (TA), (d) dissolved inorganic carbon (DIC), (e)
fugacity of carbon dioxide (fCO₂), (f) carbonate ([CO₃²⁻], (g) pH on the Total Scale, (h) aragonite saturation state
390 **(Ω_{arag}), and (i) calcite saturation state (Ω_{calc}) in the surface water. The dotted lines show the average value of the**
variabilities. This analysis is based on groups of profiles that are within 1 km apart from each other.

To present a rough estimate of the measurement uncertainties of these variables, a similar approach was used to group deep
water stations with a maximum sampling depth of >1500 m. Due to the scarcity of deep-water stations, a radius of 10 km and
395 200 m depth difference were used to find the comparison pairs. This analysis is limited to certain cruises with deep water
sampling (~5% of the data) only, thus the uncertainty estimates only hold true for these "reference" cruises, mostly with a
cruise flag of A (Table 3). They do not apply to the rest of the cruises. Results show that the DIC and TA uncertainties (0.1%
and 0.2%, respectively) are about the same as previously reported by the GLODAPv2 group (Figure 6, Table 7). Some
variables like Nitrite and Ammonium, however, suffer from uncertainties as high as ~70%. The average CTDTEMP

400   uncertainty of 0.06 °C is significantly higher than that of 0.01 °C as previously reported for the GLODAPv2 (Olsen et al.,
2020). Note the measurement uncertainties could be overestimated, because this analysis includes natural gradients due to
the large radius and depth differences, as well as any temporal changes within the 1 to 10 years (average 6 years) period.

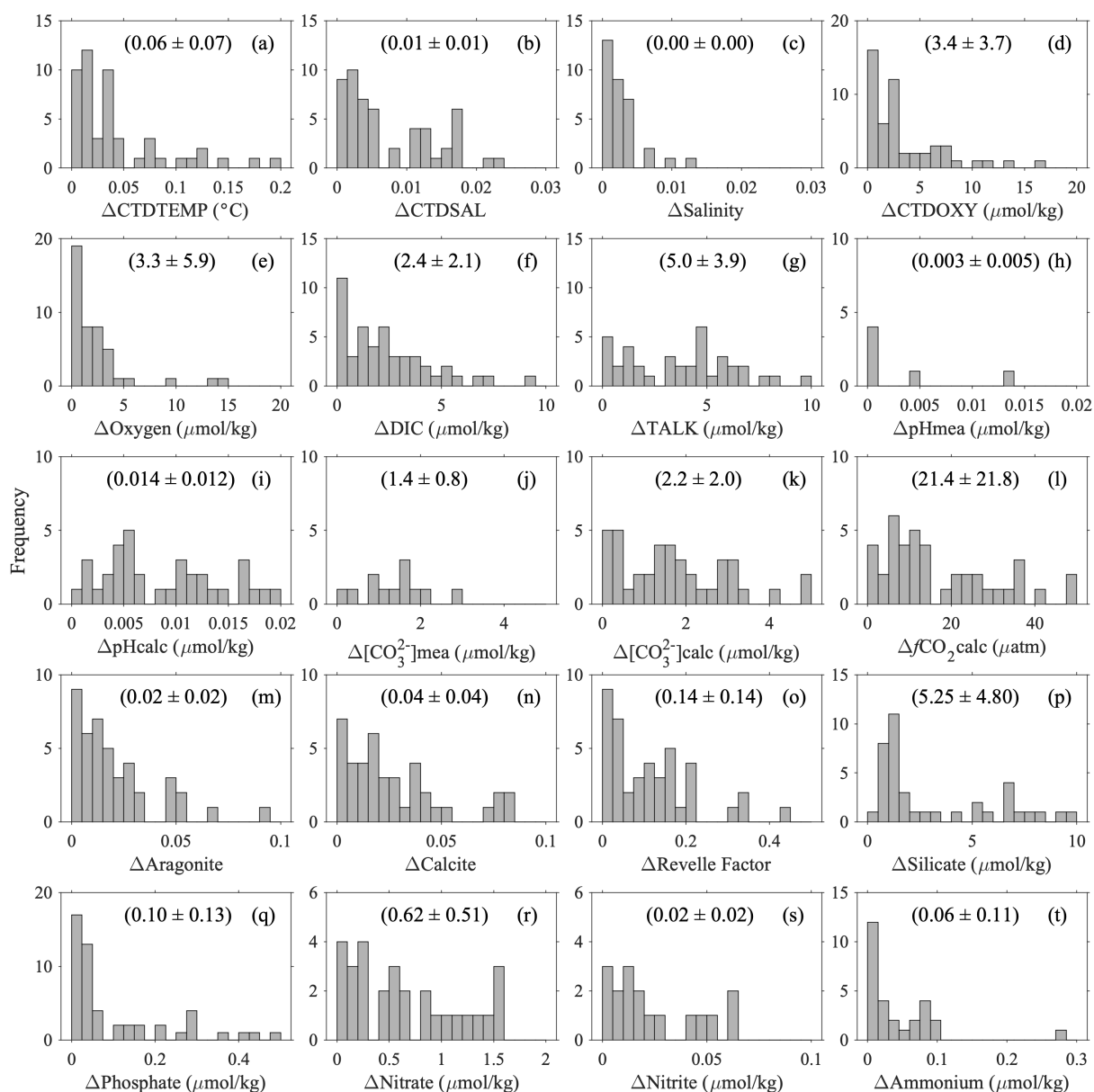



Figure 6. Uncertainties of some parameters based on deep water comparison analyses: (a) Temperature (ITS-90)
405   **measured with CTD sensors, (b) Salinity (PSS78) measured with CTD sensors, (c) discrete salinity (PSS78), (d)**
Dissolved oxygen measured with CTD sensors, (e) Dissolved oxygen concentration measured with Winkler titration,

(f) Dissolved inorganic carbon concentration, (g) Total alkalinity, (h) pH on Total Scale measured with spectrophotometers, (i) pH on Total Scale calculated from DIC, TA, and other parameters, (j) Carbonate ion concentration measured with spectrophotometers , (k-o) Carbonate ion concentration, fugacity of carbon dioxide, aragonite saturation state, calcite saturation state, and Revelle Factor calculated from DIC, TA and other parameters, (p-t) Silicate, Phosphate, Nitrate, Nitrite, and Ammonium concentrations. The values inside the parentheses are (mean values ± standard deviations).

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Table 7. Uncertainties of some variables based on an analysis that groups deep-water stations (>1000 m sampling depth) within 10 km radius, and 200 m depth difference. Stdev is the short for standard deviation. Refer to Table 2 for their

415 respective units.

Abbreviation	Mean ± stdev	Percentage	Number of pairs
CTDTEMP_ITS90	0.06 ± 0.07	-	54
CTDSAL_PSS78	0.007 ± 0.007	0.02%	53
Salinity_PSS78	0.003 ± 0.003	0.01%	33
CTDOXY	3.4 ± 3.7	4%	51
Oxygen	3.3 ± 5.9	3%	47
DIC	2.4 ± 2.1	0.1%	48
TALK	5.0 ± 3 .9	0.2%	45
pH_TS_insitu_measured	0.003 ± 0.005	0.04%	6
pH_TS_insitu_calculated	0.01 ± 0.01	0.2%	44
Carbonate_insitu_measured	1.4 ± 0.8	2%	12
Carbonate_insitu_calculated	2.2 ± 2.0	3%	44
fCO$_2$_insitu_measured	-	-	-
fCO$_2$_insitu_calculated	21 ± 22	3%	44
Aragonite	0.02 ± 0.02	3%	44
Calcite	0.04 ± 0.04	3%	44
Revelle_Factor	0.14 ± 0.14	1%	44
Silicate	5.3 ± 4.4	5%	50
Phosphate	0.10 ± 0.13	5%	51
Nitrate	0.6 ± 0.5	2%	29
Nitrite	0.02 ± 0.02	69%	17
Ammonium	0.06 ± 0.11	72%	29

For aragonite and calcite saturation states, their uncertainty comes primarily from the use of an empirical equation to approximate the real-world apparent solubility product (K_{sp}'). Despite the 3% number shown in Table 7, the real uncertainty of aragonite and calcite saturation states is likely >5% (Mucci, 1983; Jiang et al., 2015a; Orr et al., 2018). Best practices for

420 oceanic carbonate system calculations have been recommending the dissociation constants of Lueker et al., (2000) (Dickson et al., 2007). However, a recent study finds that in colder regions, where water temperature is < 8 °C, the constants of Lueker et al. (2000) may underestimate fCO$_2$ (a maximum difference of 55 μatm in their example), and overestimate pH and [CO$_3^{2-}$], meaning that cold ocean regions are more undersaturated than expected with respect to calcium carbonate mineral (CaCO$_3$) saturation states (Sulpis et al., 2020). This applies to a lot of Alaska coast stations. In brackish water (salinity < 20), the

425 relative uncertainty in carbonate ion concentration is worse than that in open ocean water (Dickson et al., 2007; Orr et al.,

2018). In addition, due to the way calcium concentration is derived in the CO2SYS (Riley and Tongudai, 1967; Millero, 1995), the calculated saturation states could suffer from uncertainties up to 12% for not directly measuring the calcium concentration in certain low-salinity regions (Beckwith et al., 2019; Dillon et al., 2020).

430 **7 Data availability**

The Coastal Ocean Data Analysis Product for North America (CODAP-NA) is available as a merged data product in the formats of Excel, CSV, MATLAB, and NetCDF [doi:10.25921/531n-c230, NCEI Accession: 0219960], and can be accessed with the link: https://www.ncei.noaa.gov/data/oceans/ncei/ocads/metadata/0219960.html (Jiang et al., 2020). The original

435 cruise data files have also been updated with data providers' consent and summarized in a table with the link: https://www.ncei.noaa.gov/access/ocean-acidification-data-stewardship-oads/synthesis/NAcruises.html.

8 Summary and conclusions

440 In this study, we relied on consistency checks performed in direct collaboration with the data providers who originally collected and measured the samples to QC and synthesize two decades of discrete measurements of inorganic carbon system parameters, oxygen, and nutrient chemistry data from North America's coastal oceans. The generated data product is called Coastal Ocean Data Analysis Product for North America (CODAP-NA). It is composed of 3,292 oceanographic profiles from 61 research cruises covering all continental shelves in North America (U.S. West Coast, U.S. East Coast, Gulf of

445 Mexico, and Alaska coast) from December 6, 2003 to November 22, 2018.

It is strongly recommended to measure a third carbon-related variable for consistency check purposes. The large majority of coastal OA cruises have already measured DIC and TALK, with a lot of them also measuring pH using high-precision spectrophotometric methods (Byrne and Breland, 1989; Clayton and Byrne, 1993; Dickson 1993; Liu et al., 2011; Douglas

450 and Byrne 2017). Recently, laboratories have increasingly begun to include carbonate ion concentration ($[CO_3^{2-}]$) as an additional measurable parameter of the seawater CO_2 system (Byrne and Yao, 2008; Sharp and Byrne, 2019). Uncertainty analyses suggest that cross-over adjustments could be applied to future coastal data QC. All major coastal cruises in the future are recommended to take deep water samples (>1500 m) when feasible, ideally at agreed-upon reference stations for QC purposes.

455 **Author contribution:**

All authors contributed to the writing of the paper. L-QJ coordinated with the overall effort, designed and built the tools for the QC process, participated in the QC effort, and prepared the first draft of the paper. RAF, RW, DG, LB, SA, BRC, DP, and CF contributed data to the data product, helped develop the QC strategy, provided comments on the QC tools, participated in all steps of the QC effort, and provided guidance to L-QJ on this overall effort. RW minted the new cruise

460   flags that were used by this data product. JH, CM, NM, JS, SS, and Y-YX (ranked alphabetically based on their last names) participated in the QC process. AK provided data management support to this product development. RHB, W-JC, JC, GCJ, BH, CL, JM, and JS (ranked alphabetically based on their last names) contributed data to this data product. DWT and his group measured nutrients data for the ECOA2 cruise and all of the Northeast Fisheries Science Center (NOAA/NFSC)'s Ecosystem Monitoring Program (EcoMon) cruises.

465   **Competing interest**

The authors declare that they have no conflict of interest.

Acknowledgements

Funding for this work comes from the National Oceanic and Atmospheric Administration (NOAA) Ocean Acidification Program (OAP, Project #: OAP 1903-1903). LJ is grateful to Are Olsen (University of Bergen and Bjerknes Centre for
470   Climate Research, Bergen, Norway), Robert Key (Princeton University, Princeton, New Jersey, USA), Nico Lange and Toste Tanhua (GEOMAR Helmholtz Centre for Ocean Research Kiel, Germany), and Siv Lauvset (NORCE Norwegian Research Centre, Bergen, Norway) for generously sharing their QC experiences. We thank Dr. Jon Hare (NOAA Northeastern Fisheries Science Center) for contributing nutrient data collected from U.S. Northeast coast from 2009 to 2016 (doi:10.7289/v5hq3wv3). This research was carried out in part under the auspices of the Cooperative Institute for Marine and
475   Atmospheric Studies (CIMAS), a Cooperative Institute of the University of Miami and the National Oceanic and Atmospheric Administration, cooperative agreement #NA20OAR4320472. RW, RAF, and BRC thank the National Oceanic and Atmospheric Administration's Global Ocean Monitoring and Observation division for funding the Carbon Data Management and Synthesis Grant (Fund ref. no. 100007298). We thank Nico Langue for answering multiple QC related questions along the way. LJ also wants to thank Christopher Sabine (University of Hawaii at Manoa) for the guidance and
480   encouragement from his experiences of working on the 1st version of the Global Ocean Data Analysis Project (GLODAP) long before this project was funded. We thank Maura Thomas (University of Marine) for analyzing nutrient data from some cruises, and Chris Taylor and Tamara Holzwarth-Davis (Northeast Fisheries Science Center) for help with sampling and data analysis for DIC, TALK, and DO in some cruises. This is contribution number 5161 from the NOAA Pacific Marine Environmental Laboratory.
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