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A high-frequency atmospheric and seawater pCO_2 data set from 14 open ocean sites using a moored autonomous system

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Abstract

In an intensifying effort to track ocean change and distinguish between natural and anthropogenic drivers, sustained ocean time-series measurements are becoming increasingly important. Advancements in the ocean carbon observation network over ⁵ the last decade, such as the development and deployment of Moored Autonomous pCO_2 (MAPCO₂) systems, have dramatically improved our ability to characterize ocean climate, sea-air gas exchange, and biogeochemical processes. The MAPCO₂ system provides high-resolution data that can measure interannual, seasonal, and subseasonal dynamics and constrain the impact of short-term biogeochemical variability ¹⁰ on carbon dioxide (CO₂) flux. Overall uncertainty of the MAPCO₂ using in situ calibrations with certified gas standards and post-deployment standard operating procedures is < 2 µatm for seawater partial pressure of CO₂ (pCO_2) and < 1 µatm for air pCO_2 . The MAPCO₂ maintains this level of uncertainty for over 400 days of autonomous operation. MAPCO₂ measurements are consistent with ship-board seawa-

 ter *p*CO₂ measurements and GLOBALVIEW-CO2 boundary layer atmospheric values. Here we provide an open ocean MAPCO₂ data set including over 100 000 individual air and seawater *p*CO₂ measurements on 14 surface buoys from 2004 through 2011 and a description of the methods and data quality control involved. The climate quality data provided by the MAPCO₂ has allowed for the establishment of open ocean obser vatories to track surface ocean *p*CO₂ changes around the globe. Data are available at doi:10.3334/CDIAC/OTG.TSM NDP092 and cdiac.ornl.gov/oceans/Moorings/ndp092.

1 Introduction

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The global ocean and its interactions with the atmosphere, climate, and marine ecosystem is undergoing a rapid and dramatic transition as it responds to multiple drivers on time scales from days to decades. Sustained observations guide our understanding of



this ever-evolving earth system, which in turn, informs the development of solutions for

human societies to cope with global change. The iconic Mauna Loa atmospheric carbon dioxide (CO_2) time series, or "Keeling Curve", is an example of how observations gain importance with time, as they provide the basis for understanding future changes to the earth system in the context of current and historical observations (Keeling et al.,

- ⁵ 1976; Thoning et al., 1989; Hofmann et al., 2009). Similar "ocean observatories" must be sustained in order to track ocean carbon uptake and ocean acidification in the midst of the large natural temporal and spatial variability in the marine environment. These observations will provide a record of past and current behavior of the ocean carbon system and are central to predicting its future.
- ¹⁰ While high-quality ocean carbon measurements collected on global hydrographic surveys have been carried out approximately once a decade since the 1980s, the scientific community identified that constraining ocean biogeochemical models would require much greater temporal and spatial resolution of field data. Autonomous technology to measure surface ocean carbon was developed to address this need and
- ¹⁵ has undergone rapid advancement in the last three decades (Takahashi, 1961; Weiss et al., 1982; Wanninkhof and Thoning, 1993; Feely et al., 1998; Pierrot et al., 2009). Autonomous underway systems that can measure the partial pressure of CO_2 (pCO_2) on ships were the first major breakthrough in our ability to collect high-frequency observations in the global ocean. These systems are designed to produce climate-quality data ²⁰ sets with measurements accurate to within 1 µatm for atmospheric CO_2 and 2 µatm for
- surface seawater pCO_2 . This level of accuracy has allowed the scientific community to constrain regional sea–air CO_2 fluxes to 0.2 Pg C yr⁻¹.

While underway pCO_2 observations have greatly enhanced our understanding of the spatial variability in sea–air CO_2 fluxes (Takahashi et al., 2009; Wanninkhof et al.,

25 2013), it has not solved the problem of quantifying temporal variability at a given location. Therefore, further observational improvements in these temporal dynamics are necessary, especially in highly variable regions such as the equatorial Pacific and coastal systems, in order to describe how short-term variability impacts CO₂ flux. Seawater *p*CO₂ observations that fully capture diurnal variations at a fixed site can be used



to address this need for temporal resolution and to validate annual, seasonal, and subseasonal dynamics in ocean biogeochemical models. The Moored Autonomous pCO_2 (MAPCO₂) system was developed to address this need by autonomously measuring surface ocean pCO_2 and marine boundary layer (MBL) atmospheric CO_2 every 3 h on surface buoys at approximately the same level of accuracy as underway pCO_2 systems. With this recent development of mooring autonomous pCO_2 technology, the

- combination of all three monitoring approaches (i.e., hydrographic surveys, underway, and buoy measurements) has improved our understanding of the spatial and temporal variability of ocean carbon at the sea surface. For the first time, ocean pCO_2 observations from multiple platforms have been incorporated into the most recent update of the
- Surface Ocean CO_2 Atlas (SOCAT), a data synthesis effort aimed at bringing together all available CO_2 data in the surface ocean in a common format (Bakker et al., 2014).

Here we describe the methods, data quality control, and data access for an open ocean MAPCO₂ data set collected on 14 surface buoys from 2004 through 2011. These surface ocean pCO_2 observatories are critical for characterizing the natural variability of the ocean carbon cycle, contributing to our understanding of secular trends in ocean

chemistry, validating and interpreting modeling results, and developing more sophisticated global carbon models.

2 Methods and data quality control

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In 2004, the National Oceanographic and Atmospheric Administration's (NOAA) Pacific Marine Environmental Laboratory (PMEL) began to work with the Monterey Bay Aquarium Research Institute to improve the accuracy, reliability, and ease of use of an early moored pCO₂ system developed for buoys in the equatorial Pacific. Like the well-established underway pCO₂ method (Wanninkhof and Thoning, 1993; Feely et al., 1998; Pierrot et al., 2009), this early moored system described by Friederich et al. (1995) and the MAPCO₂ system described in Sect. 2.1 combines air-water equilibrators with an infrared (IR) analyzer for CO₂ gas detection. In 2009, the MAPCO₂



technology was transferred to Battelle Memorial Institute and is commercially available as the Sealogy[®] pCO_2 monitoring system. This system is now accessible to the larger scientific community and deployed at over 50 locations in open-ocean, coastal, and coral reef environments, including on NOAA's global moored CO₂ network (www.pmel.noaa.gov/co2/story/Buoys+and+Autonomous+Systems) and Australia's Integrated Marine Observing System (imos.org.au).

2.1 Description of MAPCO₂ system

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The MAPCO₂ system includes four separate watertight cases that house the electronics, battery, transmitter, and a reference gas cylinder. The reference gases used on all
 the PMEL systems are traceable to World Meteorological Organization (WMO) standards and are provided by NOAA's Earth System Research Laboratory (ESRL). In the electronics case are the controls for the system, a memory flash card for data storage, a LI-COR LI-820 CO₂ gas analyzer, and a Sensirion SHT71 relative humidity and temperature sensor. The MAPCO₂ also includes an oxygen sensor for system diagnostic information. The LI-820 determines the CO₂ gas concentration by measuring the absorption of IR energy as a sample gas flows through an optical path. The CO₂ concentration is based on the difference ratio in the IR absorption between a reference and a sample optical path. The MAPCO₂ uses temperature and relative humidity to cal-

- culate the mole fraction of CO_2 (xCO_2) in air in equilibrium with surface seawater. The LI-820 is calibrated before every measurement using a "zero CO_2 reference" derived by scrubbing CO_2 from air using soda lime and an ESRL standard gas that spans the ocean pCO_2 values where the system is deployed. The system also includes a GPS for accurate position and time, an Iridium satellite communication link, an airblock deployed approximately 1 m above the ocean surface for atmospheric sampling, and an
- ²⁵ "h" shaped bubble equilibrator assembly described by Friederich et al. (1995). The equilibrator is the only part of the system in seawater and is made of copper-nickel alloy to prevent bio-fouling.



A schematic diagram of the main components and sampling paths in the MAPCO₂ system is shown in Fig. 1. A typical measurement cycle, including in situ calibration and the air and seawater measurements, takes approximately 20 min. At the beginning of each cycle, the system generates a zero standard by cycling a closed loop of air through a soda lime tube to remove all of the CO₂. This scrubbed air establishes the zero calibration. Next, the system is calibrated with a high standard reference gas or "span" gas. The value of this gas is set in the MAPCO₂ system before deployment (typically ~ 500 µmol mol⁻¹). The gas flows through the detector for CO₂ analysis and is vented to the atmosphere through the airblock. Once the detector is fully flushed, the flow is stopped and the system returns to atmospheric pressure. Using a two-point calibration from the zero and span values, the LI-820 is optimized for making surface ocean CO₂ measurements.

To make the seawater *x*CO₂ measurement, the MAPCO₂ system equilibrates a closed loop of air with surface seawater in the h-shaped equilibrator, which is ¹⁵ mounted in a float designed by PMEL to ensure the optimum depth for equilibration (Fig. 2). The air cycles through the system by pumping air out of flexible polytetrafluoroethylene (PTFE) tubing to approximately 14 cm beneath the surface of the seawater. While the air bubbles through the column of water, the air comes into equilibrium with the dissolved gases in the surface seawater. This air then returns to the system, pass-

- ing through a dryer (to prevent condensation in the LI-820 detector) and the relative humidity sensor. The air then circulates through the equilibrator again. The closed loop of air repeats this cycle for 10 min. The rising air bubbles in the equilibrator create seawater circulation in the equilibrator by pushing the water up and over the horizontal leg of the equilibrator and out the short leg of the equilibrator (Fig. 2). This draws new
- water into the long leg of the equilibrator ensuring that the recirculated air is always in contact with new seawater. After 10 min of equilibration, the pump is stopped and the LI-820 values are read at 2 Hz for 30 s and averaged to give the seawater xCO_2 measurement. This is a measurement of integrated seawater CO_2 levels during the 10 min equilibration time.



After the equilibrator reading, a MBL air reading is made by drawing air in through the airblock, partially drying it, and passing it through the LI-820. Once the LI-820 path has been flushed, the flow is stopped and a 30 s average reading is collected. All measurements and calibrations are made at atmospheric pressure. The seawater 5 CO₂ measurement occurs approximately 17 min after the start of the measurement cycle followed by the air CO₂ measurement two minutes later. Response time of the MAPCO₂ is dictated by the length of the full 20 min measurement cycle, and in fast mode, the MAPCO₂ system can measure air and seawater CO₂ once every 30 min.

Different types of sensors are used throughout the system for analytical, troubleshooting, and data quality control purposes. Additional parameters measured in each cycle (i.e., zero, span, equilibrator, and air) include temperature, pressure, relative humidity, and oxygen. Other sensors can also be integrated into the MAPCO₂ including CTD (Conductivity, Temperature, and Depth) instruments with auxiliary sensors attached (e.g., dissolved oxygen, fluorescence, turbidity) and pH sensors. The raw data collected by the MAPCO₂ and integrated sensors are stored on a memory flash card, and averaged data are telemetered from the buoy via the Iridium satellite communications system. This communications system also enables the user to control the MAPCO₂ remotely. The user can determine the sampling frequency and other variables, but the MAPCO₂ is nominally designed to make CO₂ measurements every 3 h with daily data transmissions for at least 400 days.

PMEL's MAPCO₂ systems have been deployed on open ocean buoys starting in 2004 with the establishment of NOAA's global moored CO₂ network and the efforts of numerous partners (see Acknowledgements). Figure 3 illustrates the locations, number of measurements, and average ΔpCO_2 (sea–air) from the 14 surface CO₂ buoys included in this data set. These mooring ΔpCO_2 observations are consistent with results of a synthesis of underway observations reported by Takahashi et al. (2009). Other than

the Bermuda Testbed Mooring (BTM) and Japanese Kuroshio Extension Observatory (JKEO) time series, which have been discontinued, and the Multi-disciplinary Ocean Sensors for Environmental Analyses and Networks (MOSEAN) buoy, which was moved



approximately 20 km to the new Woods Hole Oceanographic Institution (WHOI) Hawaii Ocean Timeseries Station (WHOTS) location, the MAPCO₂ time series shown in Fig. 3 continue to be maintained. Seven of the 14 CO₂ buoys are located in the equatorial Pacific on the Tropical Atmosphere Ocean (TAO) array. Additional open ocean MAPCO₂ sites maintained by PMEL now exist in the North Atlantic, North Indian, and Southern oceans (see www.pmel.noaa.gov/co2/story/Buoys+and+Autonomous+Systems); however, they have been deployed since 2011 and are not included in the finalized data set presented here.

2.2 Data reduction and processing

¹⁰ Prior to deployment, each MAPCO₂ is tested in a seawater tank in the lab. PMEL maintains a MAPCO₂ system and a General Oceanics 8050 underway pCO_2 system that are permanently mounted for continuous sampling in the seawater tank. The standard MAPCO₂ is regularly compared to the underway system, which is calibrated every 8 h using four standard reference gases traceable to WMO standards. MAPCO₂ systems ¹⁵ are compared to these standard systems in the lab and proceed to deployment if measuring within 3 µmol mol⁻¹ of the standard systems. Laboratory testing of the MAPCO₂ systems suggests instrument precision is < 0.6 µmol mol⁻¹ for xCO_2 values between 100 and 600 µmol mol⁻¹.

When the MAPCO₂ is recovered from the field, the system is compared against four
 ESRL standards to verify accuracy, and the raw data stored on the internal memory flash card are downloaded to a local database. Averaged raw data are then used for final processing of each data set. Final data quality control (QC) includes implementing a post calibration curve consistent with underway methods (Feely et al., 1998). Averaged *x*CO₂(wet) seawater and air measurements (defined in Table 1) are corrected based on the linear regression between raw averaged measurements and calibration

coefficients. This process also corrects for the rare miscalibrations that occur during the deployment. Since the system is calibrated before each measurement, detector



drift, if any, is negligible. This is confirmed by a mean difference between corrected and original raw data of $-0.02 \,\mu mol \, mol^{-1}$.

Data are quality controlled and flagged according to the SOCAT guidelines (Pfeil et al., 2013). For pCO_2 mooring purposes, we use three quality flags (QF): a flag value

- ⁵ of 2 represents an acceptable measurement, 3 is a questionable measurement, and 4 is a bad measurement. A measurement can be questionable for a variety of reasons often revealed by MAPCO₂ system diagnostic information (e.g., low equilibrator pressure causing incomplete seawater equilibration), and the reasoning for each flag is included in the metadata QC log so the end user can decide whether or not to use questionable
- ¹⁰ data. Prior to a data QC software update in June 2013, xCO_2 values flagged as bad (QF = 4) were still included in the published data sets, but after the software update, bad values are replaced with -999. Other parameters published in the data sets that do not have an associated flag, such as sea surface temperature (SST) and sea surface salinity (SSS), are given a value of -999 or -9.999 when the measurement is missing or bad. As a final check of the data QC process, air $xCO_2(dry)$ data are com-
- ¹⁵ missing or bad. As a final check of the data QC process, all $xCO_2(dry)$ data are compared to MBL data from the GLOBALVIEW-CO2 product (GLOBALVIEW-CO2, 2013). In cases where air xCO_2 values from an entire MAPCO₂ deployment are offset from the GLOBALVIEW-CO2 MBL time series and the surrounding MAPCO₂ deployments, a correction (typically $\leq 3 \,\mu$ mol mol⁻¹) is applied. This correction is noted in the meta-²⁰ data and can be removed by the data user if desired.

Post-QC calculation of pCO_2 and fCO_2 (fugacity of CO_2) are made according to recommendations of the underway pCO_2 community (Pierrot et al., 2009). First, xCO_2 in dry air is calculated by

$$x \text{CO}_2(\text{dry}) = x \text{CO}_2(\text{wet}) \times \frac{P_{\text{Licor}}}{P_{\text{Licor}} - \text{VP}_{\text{Licor}}}$$
 (1)

²⁵ where xCO_2 (wet) is the LI-820 measured concentration (µmol mol⁻¹), P_{Licor} is the pressure of the air and seawater samples measured in the LI-820 (kPa) and considered atmospheric pressure, and VP_{Licor} is the vapor pressure in the LI-820 (kPa). Since the



xCO₂(wet) measurement is at equilibrium with sea surface humidity, relative humidity (RH) measurements of the air samples exiting the LI-820 are used to calculate VP_{Licor} in Eq. (1) using the following as defined by LI-COR for the IR analyzers:

$$VP_{sat} = (0.61121)(1.004)e^{\left(\frac{17.502 \times T_{RH}}{240.97 + T_{RH}}\right)}$$

$${}_{5} \text{ VP}_{\text{Licor}} = (\text{RH}_{\text{sample}} - \text{RH}_{\text{span}}) \times \frac{\text{VP}_{\text{sat}}}{100}$$

where VP_{sat} is the saturation vapor pressure of the RH sensor cell (kPa), T_{RH} is the temperature of the RH sensor (°C), RH_{sample} is the RH of the air sample (%), and RH_{span} is the RH of the span (%). We use an enhancement factor for moist air of 1.004 for 20 °C and 1000 mbar in Eq. (2) as defined by Buck (1981).

Since the MAPCO₂ equilibration occurs directly in the ocean, it does not require the warming correction necessary for underway pCO_2 systems. Therefore, pCO_2 in wet air (100 % saturation) in equilibrium with the surface seawater is calculated by

$$\rho CO_2(sat) = x CO_2(dry) \times (P_{Licor} - \rho H_2 O)$$

¹⁵ where P_{Licor} is atmospheric pressure for the air and surface seawater samples (atm) and pH_2O is the water vapor pressure (atm) at equilibrator temperature as defined by Weiss and Price (1980). fCO_2 in wet air (100% saturation) in equilibrium with the surface seawater is calculated by

$$fCO_2(sat) = pCO_2(sat) \times e^{\left[\frac{P_{\text{Licor}} \times (B_{11} + 2\delta_{12})}{R \times T}\right]}$$

where the ideal gas constant $R = 82.0578 \text{ cm}^3 \text{ atm mol}^{-1} \text{ K}^{-1}$, *T* is SST (K) from the CTD, and the B_{11} virial coefficient and δ_{12} cross virial coefficient for CO₂ as defined by Weiss (1974). The raw CO₂ data, temperature, salinity, and pressures are included in all published MAPCO₂ data sets so other data users can recalculate xCO_2 , fCO_2 , and pCO_2 . Additional parameters included with the pCO_2 mooring data set are listed and described in Table 1.



(2)

(3)

(4)

(5)

2.3 Uncertainty of *p*CO₂ measurements

Precision and accuracy of the MAPCO₂ measurements have been assessed in both laboratory and field settings. As stated in Sect. 2.2, the precision of the MAPCO₂ system in a laboratory setting is $0.6 \,\mu$ mol mol⁻¹. Standard deviation of the high-frequency

- ⁵ raw data (~ 58 repeated measurements over 30 s) in the field is a good assessment of the in situ precision of the MAPCO₂ system. Mean standard deviation of the raw data from the 14 buoy time series presented here is $0.7 \,\mu$ mol mol⁻¹ for seawater xCO₂ and $0.6 \,\mu$ mol mol⁻¹ for air xCO₂, which is similar to precision measured in the laboratory. While estimating accuracy in a laboratory setting is feasible, the more-desired estimate
- of in situ accuracy is difficult to obtain due to the limited availability of validation samples for comparison and the mismatch in space and time of these validation samples compared to the MAPCO₂ measurements. These issues related to accuracy will be discussed in more detail below. In this section, we present MAPCO₂ estimated in situ precision, accuracy, and uncertainty, which we define as the overall error of the measurement encompassing instrument precision and accuracy as well as propagation of error.

Propagation of error must be considered when calculations are based on variables with individual uncertainties. These types of errors that impact the calculated pCO_2 and fCO_2 values have been assessed for underway pCO_2 systems and are typically ²⁰ small (< 0.1 µatm) with minimal impact to the overall uncertainty when combined with the larger uncertainty (< 2 µatm) in the actual xCO_2 measurement (Feely et al., 1998; Wanninkhof and Thoning, 1993; Pierrot et al., 2009). However, we utilize a different method to calculate $xCO_2(dry)$ for the MAPCO₂ system, so it is important to address the potential error in this new method. As discussed in Sect. 2.1, the MAPCO₂ system ²⁵ uses the LI-820 and the RH to report the mole fraction of CO₂ in air in equilibrium with surface seawater, called $xCO_2(wet)$. This "partially wet" measurement typically has a RH of ~ 75 % (seawater and air samples), which is not completely dried as in the underway pCO_2 method, due to lack of drying methods available for extended



autonomous operation. However, since we measure RH and temperature of the gas stream exiting the LI-820, we can calculate $xCO_2(dry)$ using Eqs. (1)–(3). The RH measurements used to calculate $xCO_2(dry)$ have separate precisions and accuracies that can propagate through Eqs. (1)–(3) (Table 2). The total estimated precision and accuracy of $xCO_2(dry)$ is calculated by summing each variable's precision and accuracy using the root sum of squares method. As presented in Table 2, propagation of all the errors from the separate variables does not cause the precision of calculated $xCO_2(dry)$ to differ from measured $xCO_2(wet)$ and results in a small impact to the accuracy (0.1 µmol mol⁻¹).

- In addition to the propagation of error, an estimate of in situ accuracy is key to determining the overall uncertainty of the MAPCO₂ system. The GLOBALVIEW-CO2 data product maintained by NOAA ESRL can be used as one validation data set for comparison to the MAPCO₂ air xCO_2 (dry) measurements (GLOBALVIEW-CO2, 2013). Figure 4 shows 3 hourly atmospheric MAPCO₂ measurements and bi-weekly atmo-
- ¹⁵ spheric CO₂ values from the MBL layer of GLOBALVIEW-CO2 at the latitude closest to each MAPCO₂ location. Atmospheric MAPCO₂ data presented here are in the finalized, processed form as described in Sect. 2.2. Both MBL and MAPCO₂ data capture seasonal variability and long-term trends, but as expected, high-frequency MAPCO₂ measurements show short-term variability typically deviating from the smoothed MBL
- data product by < 5 µmol mol⁻¹ (Fig. 4). The Mauna Loa atmospheric CO₂ record is also shown in Fig. 4c and provides a reference for illustrating the larger seasonal variability in the lower atmosphere directly influenced by the presence of the ocean's surface. For the time series longer than 2 years, growth rates of the 3 hourly MAPCO₂ and bi-weekly MBL atmospheric CO₂ are presented in Table 3. Atmospheric CO₂ growth rates observed by five of the seven mooring time series differ from the MBL data by ≤ 0.1 µmol mol⁻¹ yr⁻¹, suggesting that the finalized MAPCO₂ observations are consistent.
- tent with other atmospheric observations collected using different methods. Further work to smooth and deseasonalize the data, as in most growth rate analyses, could



provide further insight into the source of the larger differences between the $MAPCO_2$ and MBL observed at the KEO and MOSEAN/WHOTS sites.

MAPCO₂ and MBL data are compared in more detail in Table 4. This includes descriptive statistics of the finalized, processed atmospheric data in addition to the data prior to a MBL offset if applied to a MAPCO₂ deployment. The 3 hourly MAPCO₂ measurement that is closest in time to the bi-weekly MBL estimate is used to calculate the Δ (MAPCO₂ – MBL). Pre-offset MAPCO₂ data show a slight negative bias (-1.5 ± 2.4 µmol mol⁻¹) to MBL values (Table 4). The mean difference between MAPCO₂ and MBL observations post offset are -0.3 ± 1.7 µmol mol⁻¹. Standard deviations likely reflect the natural variability in atmospheric CO₂ at the sea surface illustrated in Fig. 4. Low standard error of the mean and low confidence level values reported for

the atmospheric comparison in Table 4 suggest strong statistical significance in the mean $MAPCO_2 - MBL$ values.

Since atmospheric CO_2 is well mixed in the open ocean MBL, environmental variability introduces little error to the MAPCO₂ and MBL air comparison. The resulting mean differences in the atmospheric data are likely due primarily to uncertainty in the measurements, which in this case, we associate with the MAPCO₂ system. However, surface ocean pCO_2 exhibits large temporal and spatial variability. For example, it is common to observe variability in underway pCO_2 measurements from the R/V *Atlantic Explorer* of approximately 10 µatm within 10 km of the BTM mooring over a period of

- ²⁰ *Explorer* of approximately 10 µatm within 10 km of the BTM mooring over a period of 3 h (Fig. 5). We observe even larger variability in the eastern equatorial Pacific, with changes up to 50 µatm over a period of 3 h and > 100 µatm over the course of a day (Fig. 6a). This patchiness can create errors in comparing MAPCO₂ measurements to ship-based measurements made at safe distance from the surface buoy. In Fig. 6b,
- ²⁵ for example, the difference between the TAO125W MAPCO₂ and underway measurements from the R/V *Ka'imimoana* (made within 10 km and 10 min of the MAPCO₂ measurement) start at $\pm 2 \mu$ atm on 31 January 2006, 14:00:00 UTC, but as the ship begins to leave the surface buoy 6 h later, the measurements diverge as the MAPCO₂ starts to detect a decreasing trend in surface seawater *p*CO₂ values at the buoy location that



persists for the next 8 days. In another example shown in Fig. 6c, the 15 μ atm difference between the MAPCO₂ and underway system observed on 10 November 2008 is similar to the daily variability observed at the buoy in the 4 days prior to arrival of the *Ka'imimoana* and could reflect true differences observed by the underway and MAPCO₂ systems located 1–7 km apart. These examples highlight the difficulty of separating environmental variability and instrument uncertainty in these types of comparison exercises.

In order to minimize environmental variability while maximizing sample size for descriptive statistics, we use discrete measurements made within 10 km and 1.5 h and av-

- ¹⁰ eraged underway pCO_2 measurements made within 10 km and 10 min of the MAPCO₂ system measurements for the seawater pCO_2 comparison analysis. While underway and MAPCO₂ systems utilize similar methodology, discrete pCO_2 presented in Table 4 is calculated from measurements of dissolved inorganic carbon (DIC) and total alkalinity (TA) using the program CO2SYS developed by Lewis and Wallace (1998) with the
- ¹⁵ constants of Lueker et al. (2000). Typical error in calculated *p*CO₂ using this method is 5 %. Only finalized seawater MAPCO₂ data are used for the descriptive statistics presented in Table 4. Unlike the descriptive statistics for the MAPCO₂ air comparisons, the statistics that result from using MAPCO₂ seawater measurements pre-MBL offset are not statistically different than the finalized, post-MBL offset statistics presented in Table 4. This could be due to the large natural variability in seawater *n*CO₂ compared
- Table 4. This could be due to the large natural variability in seawater pCO_2 compared to atmospheric CO_2 .

Agreement between discrete and mooring surface ocean pCO_2 measurements is within 1.3 µatm (mean Δ in Table 4; BTM example in Fig. 5). Although more discrete measurements have been made at these and other mooring locations, this compari-

²⁵ son is based on discrete samples restricted to within 10 km and 1.5 h of the MAPCO₂ system measurements with n > 5. Even with these restrictions, it is likely that environmental variability is not completely removed and is reflected in the mean Δ standard deviations of 3.7–6.2 µatm (Table 4). The small sample sizes (\leq 10 at each site) also resulting from these restrictions create large uncertainty in mean Δ values with standard



error and confidence levels exceeding mean Δ values. This analysis shows promising results with a close agreement between discrete and MAPCO₂ measurements; however, more discrete samples will need to be collected within 10 km and 1.5 h of MAPCO₂ system measurements in order to improve the statistical significance of the seawater *p*CO₂ comparison.

Sample sizes are larger $(13 \le n \le 76)$ for the comparison between underway and MAPCO₂ measurements at the BTM, TAO125W, and TAO140W locations. While underway measurements exist at other equatorial Pacific mooring locations, comparisons within 10 km and 10 min are restricted to TAO125W and TAO140W due to the large gaps in pCO_2 mooring data, the infrequent mooring servicing ship visits to each site (~ once every 1–1.5 years), and the necessity for the mooring servicing ship to leave for the next station before the MAPCO₂ system has gone through a few cycles and measurements have stabilized. Even with these challenges, there are 76 comparison samples at BTM during the two buoy deployments in 2006–2007 (Fig. 5). These mea-

- ¹⁵ surements show a mean Δ of $1.8 \pm 4.8 \,\mu$ atm with a low confidence level values of 1.1, indicating strong statistical significance (p < 0.05) that the actual mean Δ is between 0.7 and 2.9 μ atm (Table 4). Standard deviations of the difference between the BTM vs. discrete (5.6) and underway measurements (4.8) are similar, which may be reflective of the environmental variability in this region of the surface ocean. Mean Δ in the equa-
- ²⁰ torial Pacific is higher (-3.3 ± 15.2 µatm at TAO125W and 2.1 ± 8.3 µatm at TAO140W), but statistical significance of these values is low due to the lower sample sizes and higher environmental variability in this region (Fig. 3). The largest standard deviation in mean Δ of 15.2 is at TAO125W, which is the site that exhibits the largest natural variability (i.e., total range of ~ 200 µatm, Fig. 6a) in surface seawater pCO_2 of the open ocean mooring data sets compared in Table 4.

The MAPCO₂ system has also been involved in two independent ocean pCO_2 instrument inter-comparisons. During an Alliance for Coastal Technologies (ACT) demonstration project, the difference between the MAPCO₂ system and an underway pCO_2 system was $-9 \pm 8 \mu$ atm in coastal Washington, USA waters and $-3 \pm 9 \mu$ atm in coral



reef waters of Kaneohe Bay, Hawaii, USA (Schar et al., 2010). Separating environmental variability from instrument uncertainty in this case is challenging. Small-scale environmental variability (i.e., meters) due to natural spatial patchiness of pCO_2 was determined to be 10–15 µatm at the coastal site and < 2 µatm at the coral site and may

account for much of the difference observed between the MAPCO₂ and reference measurements. An inter-comparison between buoy and underway *p*CO₂ systems held at the National Research Institute of Fishery Engineering in Hasaki, Kamisu City, Ibaraki, Japan was done in the more controlled environment of an indoor seawater pool (UN-ESCO, 2010). In this inter-comparison, the MAPCO₂ was within 1 µatm compared to the underway *p*CO₂ reference system in conditions within the calibration gas range.

In summary, the MAPCO₂ system performs very well in laboratory and field settings in comparison to a variety of other methods. Considering the precision estimate of the MAPCO₂ measurements in the field ($< \pm 0.7 \,\mu$ mol mol⁻¹), the statistically strong (p < 0.05) mean differences in MAPCO₂ vs. comparison measurements in Table 4 ($< \pm 1.8 \,\mu$ atm), and the small propagation of error resulting from the xCO₂ (dry) calculation ($< \pm 0.1 \,\mu$ mol mol⁻¹), we estimate in situ MAPCO₂ precision at $< \pm 0.7 \,\mu$ mol mol⁻¹ and accuracy at $< \pm 2.0 \,\mu$ mol mol⁻¹ for xCO₂(dry) measurements. Overall uncertainty of pCO₂ and fCO₂ observations from the MAPCO₂ system is estimated to be $< 2.0 \,\mu$ atm for values between 100 and 600 μ atm for over 400 days of autonomous operation. However, the uncertainty of finalized, quality-controlled data is likely better for atmospheric pCO₂ and fCO₂ observations at $< 1.0 \,\mu$ atm when following the post-deployment standard operating procedures described in Sect. 2.2.

3 Data description and access

Finalized MAPCO₂ data are reported to the Carbon Dioxide Information Analysis ²⁵ Center (CDIAC; cdiac.ornl.gov/oceans/Moorings) and archived at additional data centers such as the National Oceanographic Data Center (www.nodc.noaa.gov). The archived data are organized by site and deployment date. The numeric data package



(NDP) associated with this publication includes the 56 deployments listed in Table 5 and is available at doi:10.3334/CDIAC/OTG.TSM_NDP092 or cdiac.ornl.gov/oceans/Moorings/ndp092. The methods described here are associated with the mooring ρCO_2 data included in this NDP. These data are made freely available to the public and the

- scientific community in the belief that their wide dissemination will lead to greater understanding and new scientific insights. Users of the data are requested to cite this publication when using the entire open ocean mooring data set or cite according to the CDIAC data archive when using individual mooring data sets. When preparing manuscripts using these data, users are asked to invite lead *p*CO₂ mooring investigators
 - to ensure that the quality and limitations of the data are accurately represented.

The mooring data set includes 3 hourly seawater and air CO_2 observations from 14 moorings since 2004 encompassing over 100 000 individual measurements. As presented in Fig. 3, climatological means of surface ocean pCO_2 measured on moorings

- are consistent with observations from other platforms (Bakker et al., 2014; Takahashi et al., 2009); however, much of the value in high-frequency mooring observations is demonstrated at shorter time scales. Figure 3 shows that short-term (≤ 2 years) variability at the subtropical sites tends to be dominated by the seasonal cycle, and tropical sites tend to be dominated by interannual variability. At the subtropical sites, seawater
- ²⁰ CO₂ is typically highest in the summer and lowest in the winter. The Papa site is the highest-latitude mooring in this data set and exhibits approximately equal short-term variation driven by the seasonal cycle and interannual variability caused by strong weather events in this region of the North Pacific. The highest interannual variability is observed in the equatorial Pacific driven by El Niño and La Niña events (Fig. 3)
- ²⁵ and dominates any small seasonal signal that may exist in this region (Sutton et al., 2014). In the most extreme conditions, seawater pCO_2 values can vary over 100 µatm within 24 h at 0°, 125° W (Fig. 6a). Variability of 100–150 µatm is also common in the equatorial Pacific during the extension of the warm water pool during El Niño events on time scales of months and the passing of tropical instability waves on time scales



of weeks (e.g., Fig. 4 in Sutton et al., 2014). Sustained, long-term mooring time series also provide the opportunity to identify and remove the short-term variability from the time series and investigate long-term trends. For example, in a synthesis of equatorial Pacific mooring data, Sutton et al. (2014) found that the uptake of anthropogenic

⁵ CO₂ and an acceleration in equatorial upwelling since the shift in the Pacific Decadal Oscillation in 1998 has led to high rates of pCO_2 change of +2.3 to +3.3 uatm yr⁻¹ in this region. This decadal shift in CO₂ outgassing is consistent with underway pCO_2 observations made in this region since 1982 (Feely et al., 2014).

Mooring data from most of the deployments through 2010 listed in Table 5 are also included in the most recent version of SOCAT (Bakker et al., 2014). This SOCATv2.0 synthesis involves a standardized, second level quality control of 10.1 million surface seawater fCO_2 measurements from many different sources, including underway and mooring systems. SOCAT also produces a gridded surface ocean fCO_2 data product in a uniform format available at www.socat.info. Rödenbeck et al. (2013) compared the previous version of SOCAT (v1.5), which did not include mooring data, to some of the open ocean MAPCO₂ time series in Table 5. In a comparison between seawater

- pCO₂ data from the TAO170W MAPCO₂ and model estimates based on SOCATv1.5, Rödenbeck et al. (2013) find that the model estimates in the tropics are unrelated to or even opposite of the mooring observations. While the Rödenbeck et al. (2013) model captures high-frequency pCO_2 variations related to thermodynamic effects, it does not
- ²⁰ captures high-frequency pCO_2 variations related to thermodynamic ellects, it does not reproduce high pCO_2 upwelling events. We expect the recent mooring additions to SOCATv2.0 and the open ocean MAPCO₂ data set presented here to make a large impact on our efforts to model and understand the global carbon cycle in the coming years.

25 4 Conclusion

Mooring observations can play a critical role in improving our ability to model, understand, and describe the ocean carbon cycle on all time scales. In particular, time series



from remote, data-sparse areas of the ocean collected on moorings fulfill a unique niche by providing the high-resolution data necessary to explore questions about short-term variability at fixed locations. Here we provide a data set of 3 hourly surface seawater and marine boundary layer atmospheric pCO_2 observations on 14 open ocean

- ⁵ moorings in the Pacific and Atlantic from 2004 to 2011. When using the in situ and post-calibration methods described here, overall uncertainty for the MAPCO₂ data is < 2 µatm for seawater pCO_2 and < 1 µatm for air pCO_2 , making the MAPCO₂ system a climate-quality method for tracking surface ocean pCO_2 . These types of sustained, temporally-resolved observations allow us to improve our understanding of the role of
- ¹⁰ shorter-term variability and key biogeochemical processes on the global carbon system. Potential uses of these data to inform our understanding of a changing ocean include: investigating high-frequency variability in surface ocean biogeochemistry, developing seasonal CO₂ flux maps for the global oceans (e.g., Takahashi climatology and SOCAT), studying ocean acidification, and evaluating regional and global carbon ¹⁵ models.

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 that have made the MAPCO₂ systems available to the broader research and ocean observing community. The NOAA network of pCO₂ moorings would not be possible without the industrious efforts of PMEL technical and engineering staff as well as our partners and their funders that support the maintenance of the open ocean buoys: Nick Bates (BTM), Meghan Cronin (Papa and KEO), Michael McPhaden (TAO array), Al Plueddemann and Robert Weller (WHOTS and Stratus), and Uwe Send (CCE1). PMEL contribution 4061 and JISAO contribution 2254.

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Table 1. Final data variable names and descriptions.

Variable name	Description	Units	Equation (if applicable)
Mooring	mooring name as shown in Fig. 3	character string	
Latitude	average latitude during deployment	decimal degrees	
Longitude	average longitude during deployment	decimal degrees	
Date	date of measurement in UTC	MM/DD/YYYY	
Time	time of measurement in UTC	HH:MM	
xCO2_SW_wet	mole fraction of carbon dioxide in air in equilibrium with surface seawater at SST and humidity	µmol mol ⁻¹	
xCO2_SW_QF	primary flag associated with seawater xCO ₂ measurement	SOCAT standards ^a	
H2O_SW	mole fraction of water in gas from equilibrator	µmol mol ⁻¹	
xCO2_Air_wet	mole fraction of carbon dioxide in air at $\sim 1.5\text{m}$ above the sea surface at sample humidity	µmol mol ⁻¹	
xCO2 SW QF	primary flag associated with air xCO_2 measurement	SOCAT standards ^a	
H2O Air	mol fraction of water in air	umol mol ⁻¹	
Licor Atm Pressure	Atmospheric pressure at ~ 1.5 m above the sea surface	hPa	
Licor_Temp	Licor temperature	°C	
Percent_O2 ^b	$\%$ oxygen in surface seawater divided by $\%$ oxygen in air at $\sim 1.5m$ above the sea surface	%	
SST ^c	Sea surface temperature	°C	
SSS ^c	Sea surface salinity		
xCO2_SW_dry	mole fraction of carbon dioxide in dry air in equilibrium with surface seawater	µmol mol ⁻¹	1
xCO2 Air dry	mole fraction of carbon dioxide in dry air at ~ 1.5 m above the sea surface	µmol mol ⁻¹	1
fCO2_SW_sat	fugacity of carbon dioxide in wet air (100 % humidity) in equilibrium with surface sea- water	µatm	5
fCO2_Air_sat	fugacity of carbon dioxide in wet air (100 % humidity) at ~ 1.5 m above the sea surface	µatm	5
dfCO2	fCO2_SW_sat - fCO2_Air_sat	, µatm	
pCO2_SW_sat ^d	partial pressure of carbon dioxide in wet air (100 % humidity) in equilibrium with surface seawater	µatm	4
pCO2_Air_sat ^d	partial pressure of carbon dioxide in wet air (100 $\%$ humidity) at $\sim 1.5m$ above the sea surface	µatm	4
dpCO2 ^d	pCO2_SW_sat - pCO2_Air_sat	µatm	

Notes:

^a SOCAT flags used in this data set: 2 = acceptable measurement, 3 = questionable measurement, 4 = bad measurement (note: bad data values are reported in the final data file submitted to CDIAC prior to QC software upgrade in Jun 2013 but reported as -999 in files submitted after the upgrade).

^b Oxygen measured in the MAPCO₂ system is exposed to air and likely modified within the system prior to measurement. Rapid changes in oxygen are not properly captured using this method. This data should not be used as a quantitative measure of oxygen.

^c Usually measured by other academic partners at each site. See metadata for each deployment for details on SST and SSS measurements.

^d pCO₂ only presented in data sets submitted to CDIAC after Jun 2013 when QC software was upgraded to include this calculation. Data users of earlier data sets can calculate pCO₂ as defined in Eq. (4).

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Table 2. Sources of error for the calculation of $xCO_2(dry)$ at atmospheric pressure = 101 kPa, $RH_{sample} = 75\%$, $RH_{span} = 30\%$, SST = 25°C, SSS = 35, and $xCO_2(wet) = 375 \mu mol mol^{-1}$. Total estimated precision and accuracy is calculated using the root sum of squares method (RSS): RSS = $(\sum a^2)^{1/2}$.

Variable precision (±)	Effect on pre- cision of final calculation (<i>a</i>)	Variable accuracy (±)	Effect on accuracy of final calculation (<i>a</i>)
0.019 ^a	0.009	0.057 ^a	0.026
0.1 ^b	negligible	3.0 ^b	0.1
0.1 ^b	negligible	3.0 ^b	0.1
			0.052
	0.009		0.153
0.7 ^c	0.7	1.5 ^c	1.5
0.001 ^b	negligible	0.010 ^b	negligible
0.009	0.034	0.153	0.585
	0.7		1.6
	Variable precision (±) 0.019 ^a 0.1 ^b 0.1 ^b 0.7 ^c 0.001 ^b 0.009	Variable precision (\pm)Effect on pre- cision of final calculation (a)0.019a0.0090.1 ^b negligible0.1 ^b negligible0.1 ^b 0.0090.1 ^b negligible0.0090.0090.7 ^c 0.70.001 ^b negligible0.0090.0340.70.7	$\begin{array}{c c} \mbox{Variable} \\ \mbox{precision}(\pm) \\ \mbox{Precision of final} \\ \mbox{calculation (a)} \\ \end{array} \\ \begin{array}{c} \mbox{Variable} \\ \mbox{accuracy}(\pm) \\ $

Notes:

^a Error calculated using manufacturer estimated error for T_{BH} of ±0.1 °C precision and ±0.3 °C accuracy (see Eq. 2).

^b Error reported by manufacturer.

^c Precision estimate based on standard deviation of the high-frequency raw data (~ 57 repeated measurements over 1 min) in the field; accuracy estimate based on pre-deployment testing in the laboratory. Negligible indicates value < significant digits of variable.



Table 3. Growth rate of GLOBALVIEW-CO2 MBL and MAPCO ₂ atmospheric xCO ₂ time series
(GLOBALVIEW-CO2, 2013). For mooring time series locations see Fig. 3.

	Growth rate (μ mol mol ⁻¹ yr ⁻¹)			
Time series > 2 years	MBL	MAPCO ₂		
Рара	3.1	3.1		
KEO	1.7	1.1		
MOSEAN/WHOTS	2.0	1.6		
TAO125W	1.9	2.0		
TAO140W	1.9	1.9		
TAO170W	1.9	1.9		
Stratus	1.8	1.9		



Table 4. Descriptive statistics of Δ (MAPCO₂ measurement – comparison measurement). The closest MAPCO₂ value in time and space is compared to: biweekly GLOBALVIEW-CO2 MBL values from the latitude nearest to average buoy location, single discrete measurements made within 10 km and 1.5 h, and averaged underway pCO_2 measurements made within 10 km and 10 min of the MAPCO₂ system measurement. Standard error is the standard error of the mean and confidence level values illustrate the 95% probability that the actual population mean = sample mean ± confidence level.

	п	Mean	Standard Error	Standard Deviation	Confidence Level (95%)	
MAPCO ₂ air x CO ₂ (dry) comparison to MBL air (µmol mol ⁻¹)						
pre-MBL ^a offset (estimate of MAPCO ₂ system in situ accuracy)	1823	-1.5	0.1	2.4	0.1	
post-MBL ^a offset (estimate of finalized MAPCO ₂ data accuracy)	1823	-0.3	< 0.1	1.7	0.1	
MAPCO ₂ seawater pCO_2 comparison to calculated pCO_2 (µatm)	from di	screte D	IC, TA			
WHOTS vs. HOTS ^b	7	0.1	1.4	3.7	3.4	
BTM vs. BATS ^c	9	1.3	1.9	5.6	4.3	
Papa vs. Station P ^d	10	-0.4	2.0	6.2	4.5	
MAPCO ₂ seawater pCO_2 comparison to underway pCO_2 (µatm)						
BTM vs. Atlantic Explorer ^e	76	1.8	0.5	4.8	1.1	
TAO125W vs. <i>Ka'imimoana^f</i>	16	-3.3	3.8	15.2	8.1	
TAO140W vs. <i>Ka'imimoana^f</i>	13	2.1	2.3	8.3	5.0	

Notes on data sources and archives:

^a GLOBALVIEW-CO2 Marine Boundary Layer (MBL) data source: NOAA Earth System Research Laboratory, www.esrl.noaa.gov/gmd/ccgg/globalview/co2/co2_intro.html (GLOBALVIEW-CO2, 2013).

^b Hawaii Ocean Time-Series (HOTS) data source: University of Hawaii, hahana.soest.hawaii.edu/hot.

^c Bermuda Atlantic Time-series Study (BATS) data source: Bermuda Institute of Ocean Sciences, bats.bios.edu.

^d Station P data source: University of Washington and NOAA PMEL.

^e Atlantic Explorer data source: Bermuda Institute of Ocean Sciences, cdiac.ornl.gov/oceans/CARINA/.

^f Ka'imimoana data source: NOAA PMEL, cdiac.ornl.gov/oceans/VOS_Program/kaimimoana.html.



Table 5. List of open ocean mooring deployments in the open ocean $MAPCO_2$ data set. *n* is the total # measurements collected at each mooring location during these deployments.

Mooring		Start date	End date	Mooring		Start date	End date
MOSEAN/WHOTS		12/19/2004 05/29/2005 06/18/2006 01/28/2007 06/26/2007	05/23/2005 01/20/2006 12/21/2006 07/30/2007 06/05/2008	TAO110W	n	09/19/2009 03/15/2010 07/22/2010 2148	11/03/2009 07/14/2010 10/28/2010
	n	06/05/2008 07/11/2009 08/01/2010 17 645	02/12/2009 08/01/2010 07/13/2011	TAO125W		05/08/2004 03/16/2005 01/31/2006 04/13/2007 10/16/2007	12/20/2004 09/15/2005 07/08/2006 07/17/2007 11/10/2008
BTM	n	10/02/2005 07/14/2006 03/13/2007 5354	07/03/2006 03/02/2007 10/01/2007		п	11/13/2008 04/22/2010 13609	10/21/2009 11/06/2010
Рара	п	06/08/2007 06/11/2008 06/13/2009 06/16/2010 9235	06/10/2008 11/11/2008 03/27/2010 06/13/2011	TAO140W		05/23/2004 09/13/2004 03/02/2005 01/17/2006 09/14/2006 05/31/2007	09/12/2004 03/01/2005 09/22/2005 05/14/2006 12/18/2006 11/20/2007
KEO		09/28/2007 09/13/2008 09/05/2009 09/30/2010	08/08/2008 09/04/2009 09/24/2010 12/24/2010		n	05/10/2008 09/04/2009 11/26/2010 14276	09/03/2009 01/30/2010 03/23/2011
	п	9182		TAO155W	n	01/13/2010 1791	08/25/2010
JKEO	n	02/18/2007 1837	10/03/2007	TAO170W		07/04/2005	06/23/2006 08/13/2008
CCE1	2	11/11/2008 05/19/2009 12/15/2009 09/02/2010	02/06/2009 12/14/2009 09/01/2010 10/11/2010		п	08/26/2008 06/02/2009 02/03/2010 12 528	06/01/2009 12/12/2009 02/04/2011
Stratus	11	10/16/2006 10/27/2007	10/29/2007 10/27/2008	TAO165E	n	02/23/2010 2955	02/27/2011
	п	10/26/2008 01/19/2010 10889	01/18/2010 07/07/2010	TAO8S165E	n	06/22/2009 10/18/2010 6720	09/19/2010 11/15/2011

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Figure 1. Schematic diagram of main components and sampling paths within the MAPCO₂ system. The floating air-water equilibrator is shown in more detail in Fig. 2.





Figure 2. Schematic diagram of the floating air-water equilibrator assembly in the MAPCO₂ system during the seawater equilibration cycle. Air is pumped from the MAPCO₂ through a PTFE tube and bubbled into the equilibrator. As the bubbles rise through the water, the air comes into equilibrium with the dissolved gases in the surface seawater. The rising air bubbles in the equilibrator also create circulation by pushing water up and over the horizontal leg of the h-shaped equilibrator and out the short leg of the equilibrator. Image is not to scale.





Figure 3. Location of open ocean moorings in the MAPCO₂ data set. Inner circle color illustrates the mean ΔpCO_2 of the finalized data at that location. Inner circle size is relative to the environmental variability in the time series defined here as the standard deviation of seawater pCO_2 values. The outer ring shows the proportion of environmental variability in seawater pCO_2 due to the seasonal cycle (black) and interannual variability (gray). Seasonal variability is defined as the mean seasonal peak amplitude and interannual variability is the mean Δ of annual mean values. Seasonal and interannual variability cannot be quantified at JKEO with a time series of < 1 year and is represented here by an outer ring with no color.





Figure 4. MAPCO₂ and GLOBALVIEW-CO2 MBL atmospheric xCO_2 (µmol mol⁻¹) presented by latitude: **(a)** Papa MAPCO₂ (gray points) and MBL at 49° N (black line); **(b)** BTM MAPCO₂ (gray points), MBL at 30° N (black line), KEO and CCE1 MAPCO₂ (blue points), and MBL at 33° N (blue line); **(c)** MOSEAN MAPCO₂ (gray points), MBL at 24° N (black line), WHOTS MAPCO₂ (blue points), MBL at 27° N (blue line), and Mauna Loa Observatory atmospheric xCO_2 (red line); **(d)** six equatorial MAPCO₂ buoys (gray points) and MBL at 0° (black line); and **(e)** Stratus MAPCO₂ (gray points), MBL at 20° S (black line), TAO8S165E MAPCO₂ (blue points), and MBL at 9° S (blue line). MBL data from GLOBALVIEW-CO2 (2013). Mauna Loa observatory monthly mean data from Pieter Tans, NOAA/ESRL (www.esrl.noaa.gov/gmd/ccgg/ trends/) and Ralph Keeling, Scripps Institution of Oceanography (scrippsco2.ucsd.edu/).





Figure 5. Seawater *p*CO₂ values from BTM MAPCO₂ (gray points), Bermuda Atlantic Timeseries Study (BATS) discrete (plus signs), and R/V *Atlantic Explorer* underway (open circles) used in the Table 4 statistics. BATS data from Bermuda Institute of Ocean Sciences, bats.bios. edu. *Atlantic Explorer* data from Bermuda Institute of Ocean Sciences, cdiac.ornl.gov/oceans/CARINA/.





Figure 6. (a) TAO125W surface seawater MAPCO₂ observations (gray points) for the entire time series at this location with average R/V *Ka'imimoana* underway pCO_2 data within 10 km and 10 min of the MAPCO₂ measurements (black open circles). Two examples of comparison data over one week time series are shown in panels (b) and (c) with MAPCO₂ measurements corresponding to the average underway observations illustrated in gray open circles. Selection boxes in (a) are not to scale of actual axes in (b) and (c) panels. *Ka'imimoana* data from NOAA PMEL, cdiac.ornl.gov/oceans/VOS_Program/kaimimoana.html.

