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Development and validation of a DIY profiling float for indirect determination of salinity

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15 **Key words:** profiling float, low-cost, salinity, Seaglider

16 **Running header:** Salinity determination from buoyancy-driven devices

**17 Abstract**

18 Ocean salinity, a measure of salt concentration in seawater, is a key variable influencing  
19 water density and controls important processes like mixing and stratification. Temporal and  
20 spatial salinity variability is high in dynamic coastal systems like Puget Sound where salinity can  
21 range annually from 22 to 32 PSU. This variability occurs over small spatial scales and presents  
22 a challenge for obtaining high resolution salinity measurements. Salinity is typically measured  
23 using conductivity cells, a sensor that quantifies electrical current through seawater. These  
24 sensors can be deployed on ship-based CTDs, spatially-limited moorings, or costly AUVs, but  
25 can have limited long-term stability due to sensor drift and biofouling. This project assessed the  
26 possibility of salinity calculation from buoyancy-driven vehicles, culminating in the  
27 development and validation of a cost-efficient buoyancy-driven profiling float, dubbed  
28 openFloat, for the indirect determination of salinity. The openFloat measures in situ temperature  
29 and pressure through electronic sensors and determines water density at a given depth by  
30 achieving neutral buoyancy (e.g., float density equals water density) by adjusting its volume.  
31 Salinity can be determined by way of back calculation using the equation of state of seawater  
32 using these three known parameters. This proposed method of measuring salinity was tested in  
33 situ through multiple deployments of a buoyancy-driven Seaglider in Colvos Passage, WA  
34 (Puget Sound). Salinity was determined at the deepest point in the Seaglider's dive when it  
35 achieves neutral buoyancy and salinity is not recorded due to unstable flight conditions. Using  
36 Seaglider buoyancy parameters, salinity could be estimated to  $\pm 0.18$  PSU in situ, and with the  
37 openFloat salinity could be estimated to  $\pm 0.11$  PSU in the test pool, demonstrating the proof of  
38 concept needed for a more cost-effective option like the openFloat in dynamic systems where the  
39 range in salinity is large. Salinity calculation with openFloat can enable the large-scale

40 production of floats to obtain the high-resolution spatial and temporal data needed to better  
41 quantify change in dynamic coastal systems.

42

### 43 **Plain language summary**

44       The amount of salt contained in seawater, salinity, determines important ocean processes.  
45 Salinity changes by location and over different time periods. Salinity changes a lot over small  
46 distances in dynamic coastal regions like Puget Sound, WA. Salinity is typically measured with  
47 high-accuracy electronic sensors that are used from ships or ocean robots. While these sensors  
48 are widely used, they can be costly and have technical limitations the longer they are deployed.  
49 This project therefore developed and validated a cost-efficient buoyancy-driven profiling float,  
50 dubbed openFloat, for cheaply measuring salinity in dynamic coastal systems where high  
51 accuracy data is not needed to observe changes. The concept was first validated by performing  
52 the salinity estimation with an industry-grade ocean robot and then by performing the estimation  
53 with the openFloat in a test tank. The estimation of salinity through the development of the  
54 openFloat would enable the large-scale production of ocean robots to observe large changes in  
55 salinity between nearby locations in dynamic coastal systems such as Puget Sound, WA, with its  
56  $\pm 0.11$  PSU estimation capabilities.

## 57 **Introduction**

58           Ocean salinity, a measure of salt concentration in seawater, is a key variable influencing  
59 water density and controls important processes, like mixing and stratification. For most of  
60 oceanographic history, salinity was determined by taking the ratio of dry salt to evaporated water  
61 (e.g., grams of salt per kilogram of evaporated seawater) or through Knudsen titrations (Cox,  
62 1963). These methods require many liters of bottle samples to be analyzed at sea or stored and  
63 transported for laboratory analysis. Knudsen titrations are historically prone to error due to  
64 chemical leaching between the samples and the glass storage vessels (Cox, 1963) during long  
65 transportation periods to land-based laboratories before ship-based lab procedures were common.  
66 More recently, bottle samples to determine high accuracy salinity have been used on cruises  
67 (e.g., GO-SHIP), but these methods can be cost inefficient, as they require the use of a ship to  
68 collect bottle samples for these measurements. Alternatively, the innovation of salinity sensors  
69 that use conductivity – a measure of electrical current through a material – as a proxy for salt  
70 concentration in seawater enabled the measurement of salt concentration with circuitry.  
71 Conductivity is used as a proxy for salinity because a well-defined positive correlation exists  
72 between water conductivity and salt concentration (Cox, 1963; Woody et al., 2000).

73           Initial seawater conductivity sensors were benchtop instruments and could not be used in  
74 situ, requiring bottle samples to be obtained from various depths for ship-board analysis. The  
75 advent of waterproofing technologies enabled in situ conductivity measurements to be made with  
76 conductivity cells. In situ salinity measurements with conductivity cells do not require bottle  
77 samples to be obtained regularly, but can be prone to error due to sensor calibration drift over  
78 time and biofouling (Woody et al., 2000). Salinity error due to sensor drift is an important  
79 consideration, particularly in the homogeneous deep ocean where high levels of data accuracy

80 (within +/- 0.0003 S/m) are required to differentiate between nearly identical water parcels.  
81 Conductivity cells capable of deep ocean salinity measurements undergo extensive regularly  
82 scheduled calibration and are expensive due to their high level of data accuracy. Common  
83 methods for measuring deep ocean salinity are through the use of autonomous underwater  
84 vehicles (AUVs) such as Seagliders or Argo floats equipped with conductivity cells. The  
85 accuracy of these sensors is optimal for deep or open ocean environments that experience low  
86 variability (e.g., Hawaii: 34 – 35 PSU ) (e.g., Lukas and Santiago-Mandujano, 2008; Stammer et  
87 al., 2021). However, the level of accuracy achieved by these sensors is often not needed to  
88 quantify variations in dynamic coastal estuarine locations such as Puget Sound, WA, where  
89 temporal and spatial variability of salinity is relatively large (e.g., 20 – 33 PSU) (e.g., Moore et  
90 al., 2008; MacCready et al., 2021).

91         Arrays of salinity sensors offer better data resolution to characterize the spatial and  
92 temporal variability of salinity seen in dynamic coastal systems. However, high-cost sensors can  
93 be prohibitive in these ecosystems due to the need for many sensors to observe their high  
94 variability. Current shipboard and robotic salinity sampling techniques obtain high accuracy  
95 data, but are limited in that they cannot be deployed broadly to encompass a high-variability  
96 region like Puget Sound due to cost. The Argo float fleet, an array of buoyancy-driven profiling  
97 floats, achieves a similar goal on a global scale for deep ocean environments, but are not suitable  
98 for shallower coastal areas. This study sought to develop a low-cost DIY profiling float – the  
99 openFloat – capable of salinity estimation to address this need. The main requirement for the  
100 development of the openFloat was for it to be inexpensively reproducible to increase the possible  
101 spatial resolution of observations needed to characterize dynamic systems through mass  
102 production. This study built a DIY buoyancy-driven profiling float to determine salinity through

103 back-calculation of the TEOS-10 equation of state of seawater (McDougall et al., 2012) by  
104 inexpensively measuring temperature and pressure along with known properties of the float (e.g.,  
105 density at neutral buoyancy). This concept was first tested with an industry-grade AUV called a  
106 Seaglider.

107         Seagliders are another type of buoyancy-driven AUV that profile the water column with  
108 similar methods to Argo floats and the proposed openFloat. They can collect data between  
109 designated GPS coordinates by flying through the water in elongated V-shaped dives. Seagliders  
110 produce thrust to propel through water by adjusting their volume to become more or less dense  
111 than the surrounding water; the sinking or floating motion of the Seaglider in combination with  
112 its wings produces forward thrust, much like an airplane. Many Seagliders do not have  
113 conductivity sensors that actively pump water through itself and rely on near-laminar flow over  
114 its outer shell to achieve quality measurements. Seagliders experience laminar flow over the  
115 outer shell during all stages of a dive cycle except for the transition between diving and  
116 climbing. Consequently, Seagliders do not measure salinity for the deepest five to ten meters of  
117 any dive. This does not prove to be an issue when Seagliders are flown in water significantly  
118 deeper than its intended maximum depth (e.g., the deep ocean). In shallower water, such as Puget  
119 Sound, Seagliders are often flown to a maximum depth of ~10 m above the seafloor. This can  
120 result in a lack of salinity data for one-tenth or more of sampled depths.

121         This project addressed the question: How well can salinity be estimated from buoyancy-  
122 driven vehicles using vehicle density parameters? First, the salinity back-calculation method was  
123 validated in situ by performing the same calculations with data from Seaglider SG175, then the  
124 same method was used to calculate salinity with the openFloat in a test pool. SG175 was  
125 deployed twice in Colvos Passage, WA (Puget Sound), a channel that experiences high levels of

126 turbulent mixing due to sills – thin areas of rapidly reduced depth – in the local bathymetry  
127 (Seim and Gregg, 1997; Ovall, 2019). Flood tides bring water into Southern Puget Sound  
128 through East Passage (the eastern edge of Vashon Island) where it is then upwelled and mixed  
129 over a sill in the Tacoma Narrows; the ebb tide then again moves water through the Tacoma  
130 Narrows and into Colvos Passage (Seim and Gregg, 1997; Ovall, 2019). The high levels of  
131 mixing around Colvos Passage cause spatial variability of salinity to be low, enabling high  
132 confidence that calculated salinity will match measured salinity. I tested the hypothesis that  
133 salinity can be estimated to an accuracy suitable for highly variable coastal environments using  
134 buoyancy-driven devices (Seaglider and openFloat) when neutrally buoyant from the TEOS-10  
135 equation of state of seawater if in situ temperature and pressure are measured.

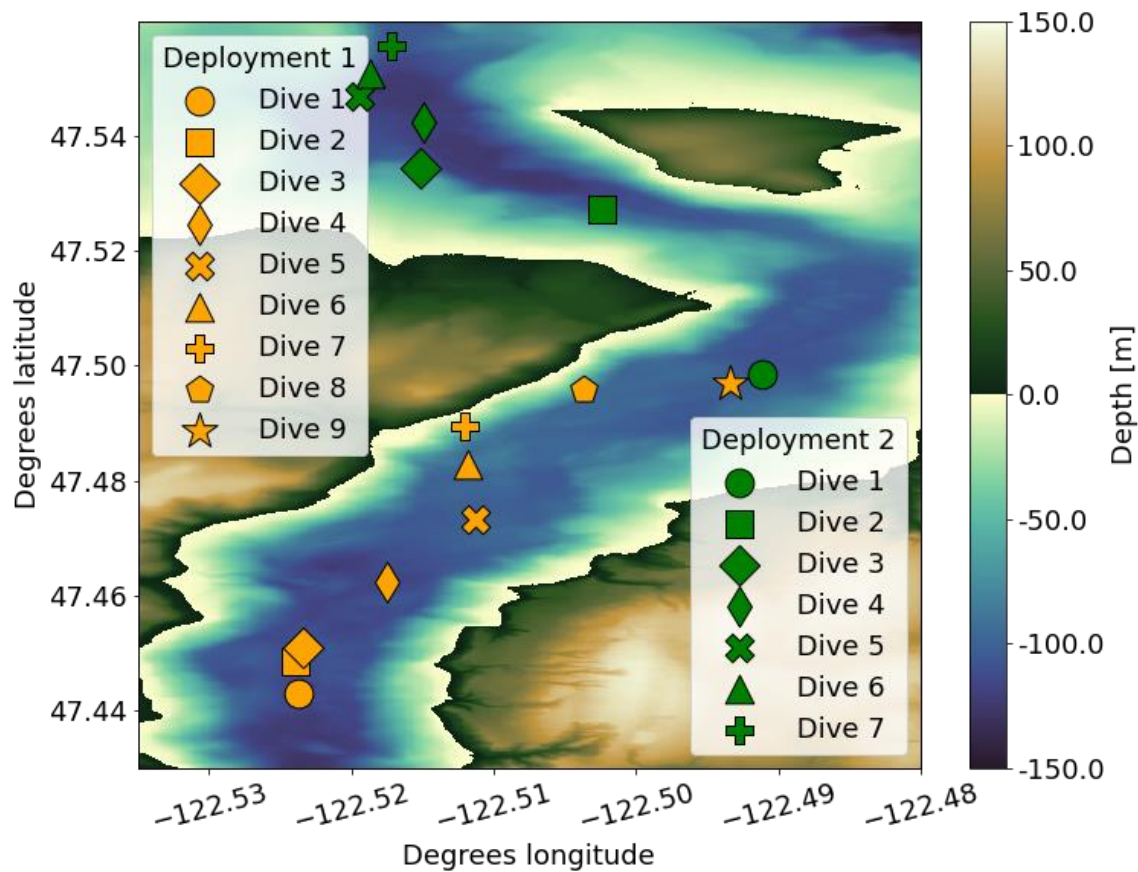
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## 137 **Methods**

### 138 *Colvos Passage Seaglider deployments*

139 Seaglider SG175 was deployed in Colvos Passage, WA, on 15 December 2023 from the  
140 R/V Weelander and completed nine dives between 6:30 PST and 14:30 PST. SG175 was again  
141 deployed in Colvos Passage on 09 February 2024 and completed eight dives between 8:30 PST  
142 and 3:30 PST. The Seaglider was deployed and recovered over the side of the research vessel  
143 using the stock cradle delivered with all Seagliders. SG175 was equipped with a standard  
144 Seabird Scientific conductivity and temperature (CT) sail and Keistler pressure sensor.

145 The first deployment began at the southern end of Colvos Passage and ended near Blake  
 146 Island, WA (Figure 1). The second deployment began near the northern end of Colvos Passage  
 147 and ended north of Blake Island (Figure 1). On average, the Seaglider dove to ~90 m with dive  
 148 times ranging between 30 and 60 minutes and vertical velocities set to be  $\pm 10 \text{ cm s}^{-1}$  depending  
 149 on ascent or descent. Each dive covered one to two kilometers in lateral distance. This Seaglider  
 150 deployments served to collect data as a proof of concept for back-calculating salinity from the  
 151 known density of an AUV at neutral buoyancy (e.g., AUV density is equal to seawater density),  
 152 in situ temperature, and pressure, while the openFloat was still in development. The Seaglider  
 153 did not collect salinity data at apogee (the bottom of the dive path) presenting an opportunity to  
 154 test the accuracy and applicability of the salinity calculation in lieu of direct sensor



**Figure 1.** Dive locations of SG175 in Colvos Passage, WA, during the 15 December 2023 (yellow) and 9 February 2024 (green) deployments. The net northward movement is both due to planned waypoints and a northward outgoing tide.

155 measurements. Colvos Passage experiences high levels of mixing where water parcels  
156 throughout the channel are nearly identical, facilitating higher confidence that proof-of-concept  
157 salinity calculations would not be influenced by high environmental variability.

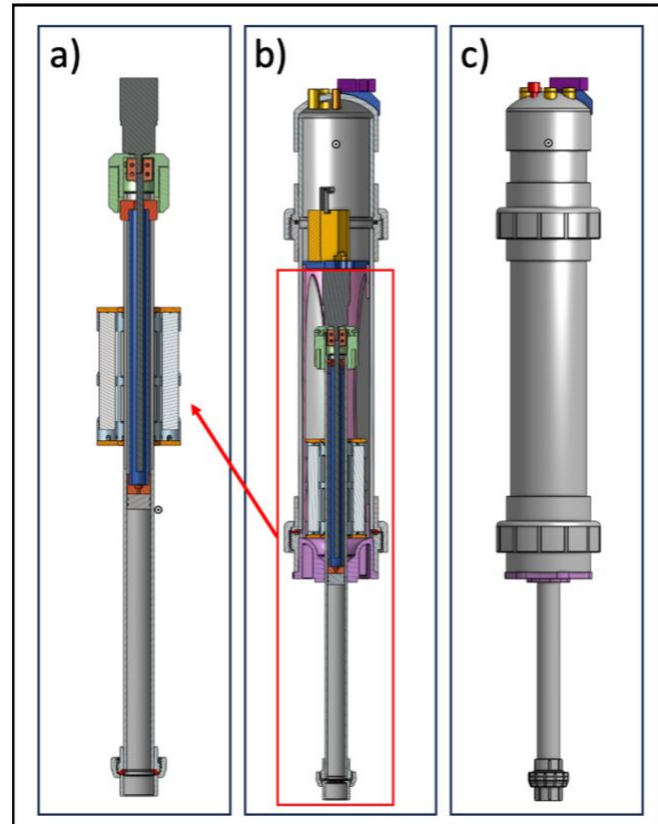
158         The salinity back-calculation was performed by feeding in situ temperature ( $^{\circ}\text{C}$ ) and  
159 pressure (dBar), and Seaglider density ( $\text{kg m}^{-3}$ ) at neutral buoyancy to the Gibbs Seawater  
160 MATLAB package function named “gsw\_SA\_from\_rho\_t\_exact.” This function calculated  
161 absolute salinity given in situ temperature, pressure, and density. Seagliders are designed to cross  
162 neutral buoyancy during their apogee maneuver (Eriksen et al., 2001). The Seaglider’s density at  
163 apogee was calculated by first isolating the base volume of the Seaglider (neglecting any added  
164 volume of the external buoyancy bladder) by subtracting the maximum volume of the buoyancy  
165 bladder from the maximum volume of the Seaglider (e.g.,  $V_{base} = V_{glider_{MAX}} - V_{bladder_{MAX}}$ ,  
166 where V is volume). This volume includes all external components of the Seaglider except for  
167 the buoyancy bladder. The volume of the buoyancy bladder at the maximum depth of apogee was  
168 then added to the Seaglider’s base volume to find the Seaglider’s total volume during apogee  
169 (e.g.,  $V_{glider_{apogee}} = V_{glider_{base}} + V_{bladder_{apogee}}$ ). The Seaglider’s density during apogee was  
170 then computed by dividing the Seaglider’s known mass by its volume at apogee.

171 ***Building the openFloat***

172 The openFloat (Figure 2) was built using  
 173 stock hardware, common electrical sensing,  
 174 and 3D-printed components. All 3D-printed  
 175 components were 3D printed with PETG  
 176 plastic using Prusa MK3s+ and Anycubic  
 177 Kobra2 Max 3D printers with suggested  
 178 print settings by the printer manufacturers.  
 179 Structural components, such as the motor  
 180 mount, were printed with 100% infill  
 181 density and five walls while non-structural  
 182 components, such as electrical board  
 183 mounts, were printed with 50% infill density  
 184 and three walls.

185 This openFloat project was built in the

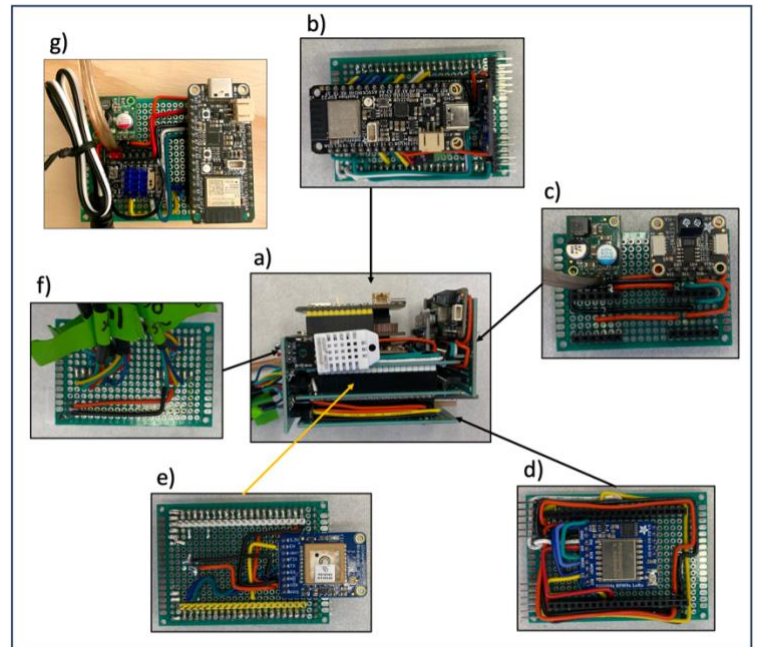
186 following five stages: electronics and sensors, variable buoyancy engine (VBE), pressure hull  
 187 and sensor cap, sensor and motor control code, and sensor calibration. The openFloat electronics  
 188 (Figure 3) are controlled by two ESP32 Feather Huzzah V2 microcontrollers paired via  
 189 Bluetooth. The first ESP32 (Figure 3a and b) controls all onboard sensors and writes associated  
 190 data to a CSV file on the internal SD card while also sending state changes to the second ESP32.



**Figure 2.** a) Isolated image of the buoyancy engine with the motor and motor mount on top, the battery pack (middle), and oil reservoir (bottom). b) Internal cross-section of the openFloat showing the internal 3D printed components (all colored components) used to hold the electrical system and operate the buoyancy piston. c) External side view of the openFloat showing the sensor cap (top), PVC pressure casing (grey components), and buoyancy engine (bottom).

191 These state changes control the second ESP32 to adjust the VBE, effectively making the float  
 192 move vertically through the water column. The data sensing electronics include a micro-SD card  
 193 reader (Figure 3b), real-time clock (Figure 3b), electrical current meter (Figure 3c), radio for data  
 194 telemetry (Figure 3d), GPS (Figure 3e), three different temperature sensors (Figure 3f), pressure  
 195 (Figure 3f), and light (Figure 3f) are mounted on custom circuit boards with through-solder

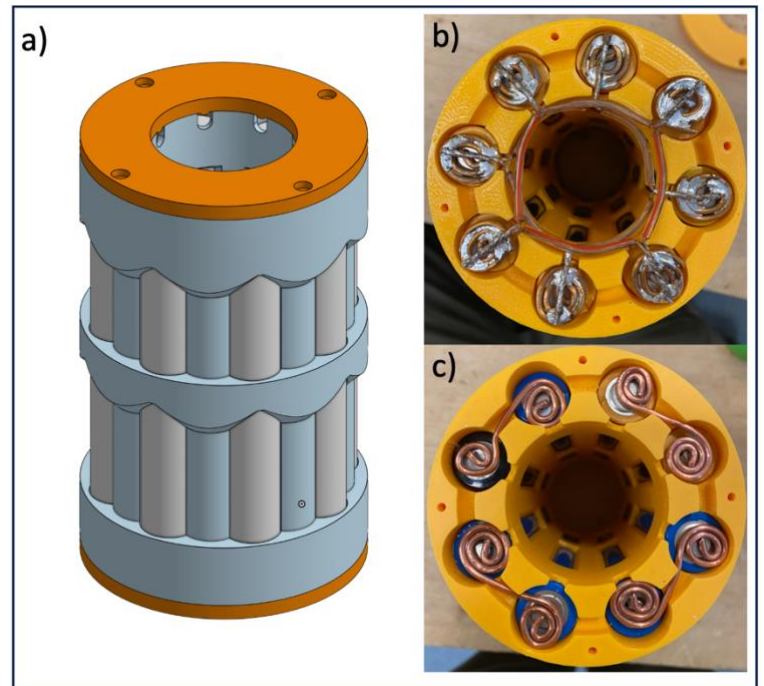
196 connections (Figure 3a). The second ESP32  
 197 (Figure 3g) controls the VBE piston position  
 198 and receives movement commands over  
 199 Bluetooth during state changes of the  
 200 openFloat (e.g., changing from sinking to  
 201 floating). The pressure (MS580305\_BA) and  
 202 temperature sensors (DS18B20 and built-in  
 203 sensor in the MS5803) chosen for the  
 204 openFloat are commonly used for DIY ocean  
 205 sensors (Lyman et al., 2020; Lauer et al.,  
 206 2020). The main electronics (Figure 3a-f)  
 207 are stacked vertically via male-female header  
 208 pins to minimize the cross-sectional area they  
 209 occupy within the openFloat's pressure hull.  
 210 External sensors were waterproofed and  
 211 potted in the openFloat's top cap with epoxy,  
 212 enabling through connections of wires from the outside of the openFloat to the main board.



**Figure 3.** a) Side profile of the assembled electronics mainboard with all electrical connections made. b) Top view of the microcontroller board (board 1) which contains the internal pressure and temperature sensor, 3-axis accelerometer and gyroscope (compass), real-time-clock, and micro-SD card reader. c) Top view of the main power board (board 5) with a 5 V step-down converter and precision electrical current meter. d) Top view of the data telemetry board (board 3) equipped with a 915 MHz radio chip. e) Bottom view of the GPS board (board 2). f) Top view of the sensor signal distribution board (board 4) which directs signals from the external pressure, temperature, lux to the microcontroller board; this board also holds the internal humidity sensor. g) Top view of the motor control board equipped with a microcontroller and stepper motor controller.

213 The VBE (Figure 2a) is built from half-inch and one-inch PVC pipe, a TR8x8 ACME  
 214 threaded rod, a NEMA 17 stepper motor with a 27:1 gear reduction, and 3D printed connectors.  
 215 The VBE changes the openFloat's volume by actuating a piston to displace varying amounts of  
 216 seawater (0 to ~250 mL). Pressure hull (Figure 2b and c) components are comprised of four-inch

217 PVC pipe and associated connectors (see  
 218 Appendix Table 1 for complete list). The  
 219 main body (Figure 2c) of the pressure hull is  
 220 a one-meter length of four-inch PVC with  
 221 flange union connectors on either end. These  
 222 flanges have built-in O-rings and enable the  
 223 float to be opened to disconnect electronics,  
 224 retrieve data from the on-board SD card, and  
 225 remove batteries for recharging. Both the top  
 226 sensor cap and bottom four-inch to one-inch



227 connector are filled with epoxy. The top  
 228 sensor cap was filled with epoxy to ensure an  
 229 air-tight seal around the 3D-printed cable  
 230 glands that are inserted through bore holes in

**Figure 4.** a) CAD model of the battery pack to hold sixteen 18650 batteries; blue and orange components were 3D-printed, while the grey rods represent the batteries. b) Star-patterned wiring to connect groups of four 18650 batteries in parallel. c) Wiring to connect groups of four 18650 batteries in series. b) and c) are on opposite end of the yellow 3D-printed battery pack. The battery pack is connected to the electronics through common 18 AWG speaker wire.

231 the sensor cap; the bottom four-inch to one-inch connector was filled with epoxy both to lower  
 232 the float's center of mass and ensure the VBE was solidly mounted. During this stage of the  
 233 project, the openFloat's battery pack (Figure 4a) was also constructed. The battery pack contains  
 234 sixteen 18650 Li-ion batteries (3.7 V each when fully charged) grouped in four parallel groups  
 235 (Figure 4b) of four batteries in series (Figure 4c). This resultant battery is ~15 V with ~14,400

236 mAh of charge and can power the openFloat conservatively for five hours. These batteries are  
 237 contained in a 3D-printed case (Figure 4a-c) and fit around the outer pipe of the VBE (Figure  
 238 2a).

239 All electrical components of the openFloat are programmed using the open-source Arduino  
 240 IDE and make use of Adafruit, stepper motor, Bluetooth, and clock libraries (the full repository  
 241 of code can be found here: [https://github.com/cflaim1123/openFloat\\_thesis/tree/main](https://github.com/cflaim1123/openFloat_thesis/tree/main)). The  
 242 openFloat runs as a state machine to control its dive cycle. State machines are blocks of  
 243 computer code that have finite quantities of Boolean states that can be toggled as True or False.  
 244 When specific states are True, other blocks of code are triggered to run, causing the VBE to  
 245 increase buoyancy, for example. The openFloat continually runs two state machines to create the  
 246 dive cycle. The first state machine tracks if the openFloat is in functioning order

247

248 **Table 1.** List of openFloat states used by its state machine operating system.

Machine	State	Description
Dive cycle controller	Dive	When True, the VBE performs a bleed move where oil is removed from the external bladder, causing the float to descend.
	Climb	When True, the VBE performs a push move where oil is added to the external bladder, causing the float to ascend.
	Park	When True, the VBE maintains its position waiting for the float to become neutrally buoyant.
	Surface	When True, the float maximizes its volume when already at the surface to increase the antennas' height to increase chance of successful data transmission and GPS fix.
Force surface controller	Abort	When True, all currently True states of the float are set to False, sampling is stopped, and the VBE performs a push move to maximize the float's volume, causing it to perform an emergency surface maneuver.
	All clear	When True, the openFloat is safe to proceed as normal and will continue diving until its dive duration requirements are met.

249

250 (e.g., if internal pressure or humidity changes, this state machine forces the openFloat to  
251 surface); the second state machine controls when and how much the VBE changes the  
252 openFloat's density. The main states of the openFloat and their functions are described in Table  
253 1. The openFloat also has global parameters to determine properties such as sampling rate or  
254 deep sleep frequency that are altered by each state.

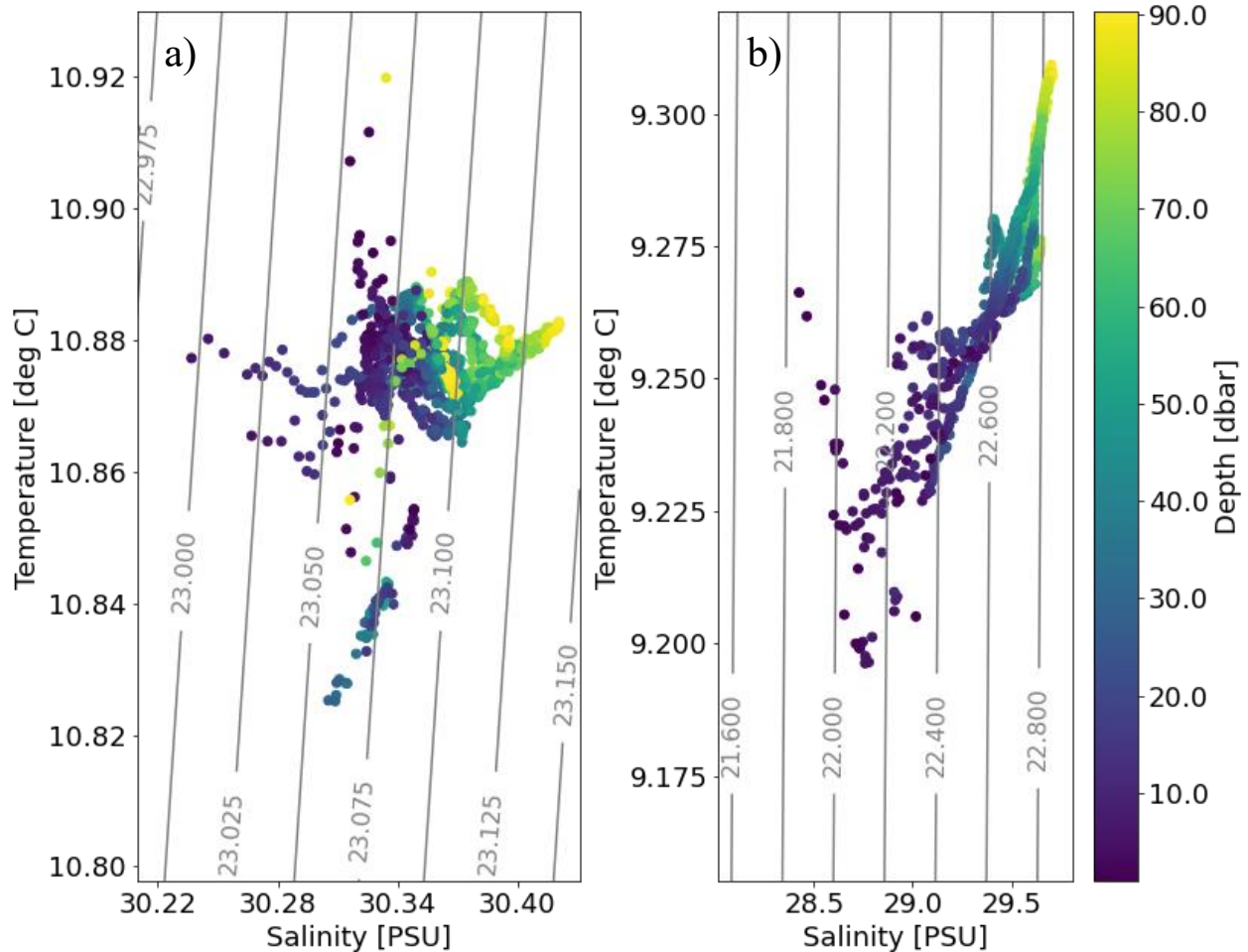
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### 256 *Testing the openFloat*

257 The openFloat was tested in a four-meter-deep saline test tank in the Ocean Sciences  
258 Building at University of Washington Seattle campus. The openFloat was made neutrally  
259 buoyant by adding mass (~3 kg) to the pressure hull in the form of lead strips until it held  
260 position at one meter water depth. A YSI EXO Sonde CTD was then used to find the density of  
261 the test tank when the openFloat was neutrally buoyant. Since the openFloat was neutrally  
262 buoyant (e.g., float density equals tank density) and of known mass, its volume was calculated  
263 by dividing the float's mass by the test tank density (e.g.,  $V_{float} = \frac{M_{float}}{\rho_{tank}}$ , under the assumption  
264  $\rho_{tank} = \rho_{float}$ ). This calculated volume was later used in salinity calculations.

265 Test tank salinity was determined by taking pressure and temperature readings every two  
266 seconds for five minutes from the neutrally buoyant openFloat. The openFloat was collocated  
267 with a CTD at one meter for direct measurement comparison. An offset calibration was applied  
268 to the openFloat's temperature readings by subtracting the difference in mean temperature from  
269 the float and CTD from the float's measured value (e.g.,  $T_{corrected} = T_{float} - avg(T_{float} -$   
270  $T_{CTD})$ ). This was done to ensure spurious salinity measurements were not made due to  
271 uncalibrated temperature measurements from the MS5803 pressure sensor. In situ pressure,  
272 corrected temperature, and calculated float density were then fed into the same MATLAB

273 “gsw\_SA\_from\_rho\_t\_exact” to calculate salinity with units of  $\text{g kg}^{-1}$ . This value was then  
 274 converted to match the PSU unit used by the Seaglider and CTD using the Gibb’s Seawater  
 275 function “SP\_from\_SA” to not introduce error into the comparison values. The openFloat’s  
 276 salinity estimation was then intercalibrated with the YSI EXO Sonde CTD using the same



**Figure 5.** In situ Temperature plotted against salinity measured by SG175 with isopycnals super-imposed showing the clustering of density measurements during the a) 15 December 2023 Colvos passage b) 9 February 2024 Colvos passage deployments. Depth is shown as color.

277 method applied to the temperature values (e.g.,  $S_{corrected} = S_{float} - avg(S_{float} - S_{CTD} * k)$ ,  
 278 where  $k$  is a conversion factor to account for the difference in conductivity readings between test  
 279 tank water and seawater) to yield corrected values.

280

**281 Results**

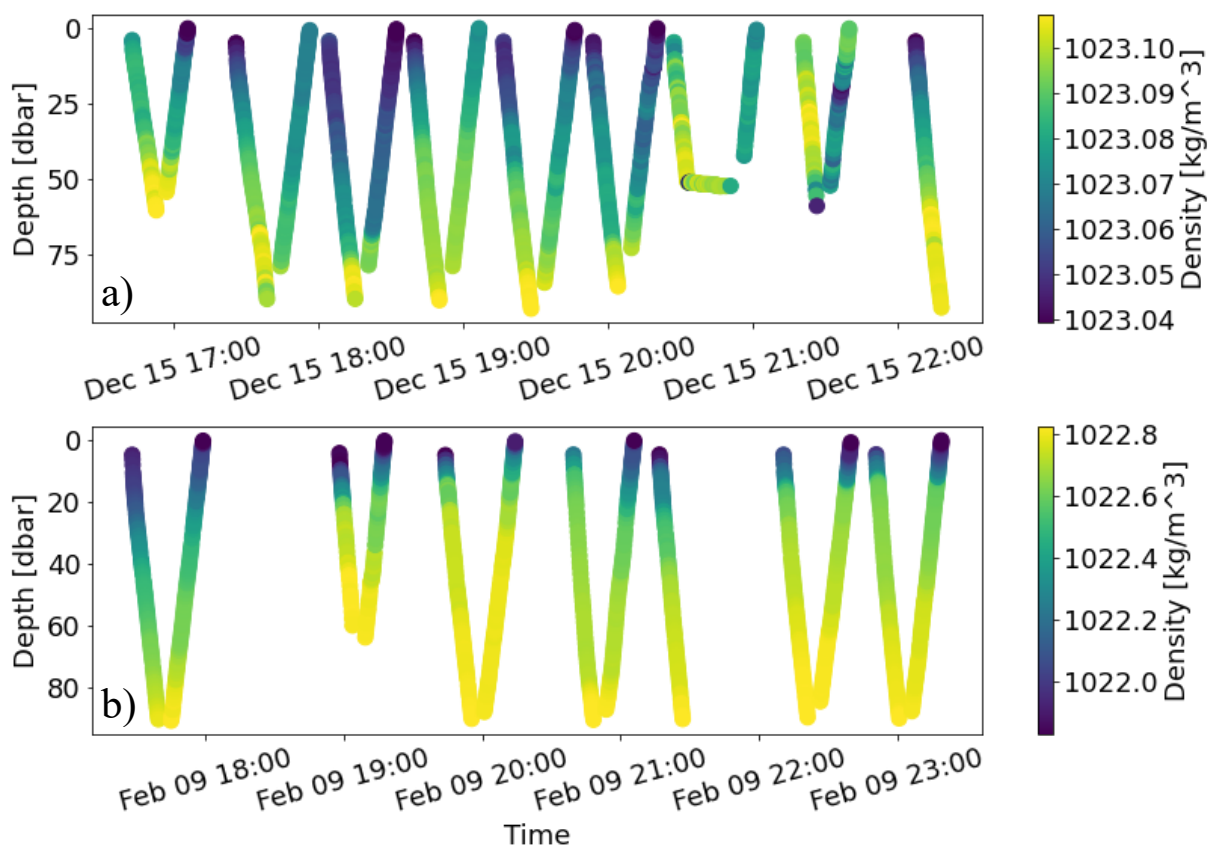
282           The first and main result of this study is the development of the electrical and mechanical  
283 systems that yield a cost-effective DIY profiling float capable of the indirect determination of  
284 salinity. This cost of the openFloat (~600 USD) is one to three orders of magnitude cheaper than  
285 existing buoyancy-driven systems (e.g.,  $\mu$ floats (micro floats) (Harrison, 2021), Argo floats,  
286 Seagliders) and enables large-scale production due to reduced production cost. This study  
287 additionally found that salinity can be determined from buoyancy-driven AUVs (commercial and  
288 DIY) if they are of known mass and volume and achieve neutral buoyancy while measuring in  
289 situ temperature and pressure.

290

**291 *Seaglider salinity calculations***

292           Colvos Passage, WA, was well-mixed on 15 December 2023 during the first Seaglider  
293 deployment. In situ temperature and measured salinity varied between 10.87 and 10.92 °C and  
294 30.22 and 30.40 PSU, respectively (Figure 5a). Sigma theta varied by less than 0.3 kg m<sup>-3</sup> during  
295 all dives (Figure 5a). Sigma theta was between 23.02 and 23.06 at the surface while sigma theta  
296 at depth was between 23.1 and 23.12. The water column was more stratified and fresher on 09  
297 February 2024 (Figure 5b) during the second deployment enabling the salinity back-calculation  
298 test across a wider range of environmental conditions. Temperature ranged between 9.175 and

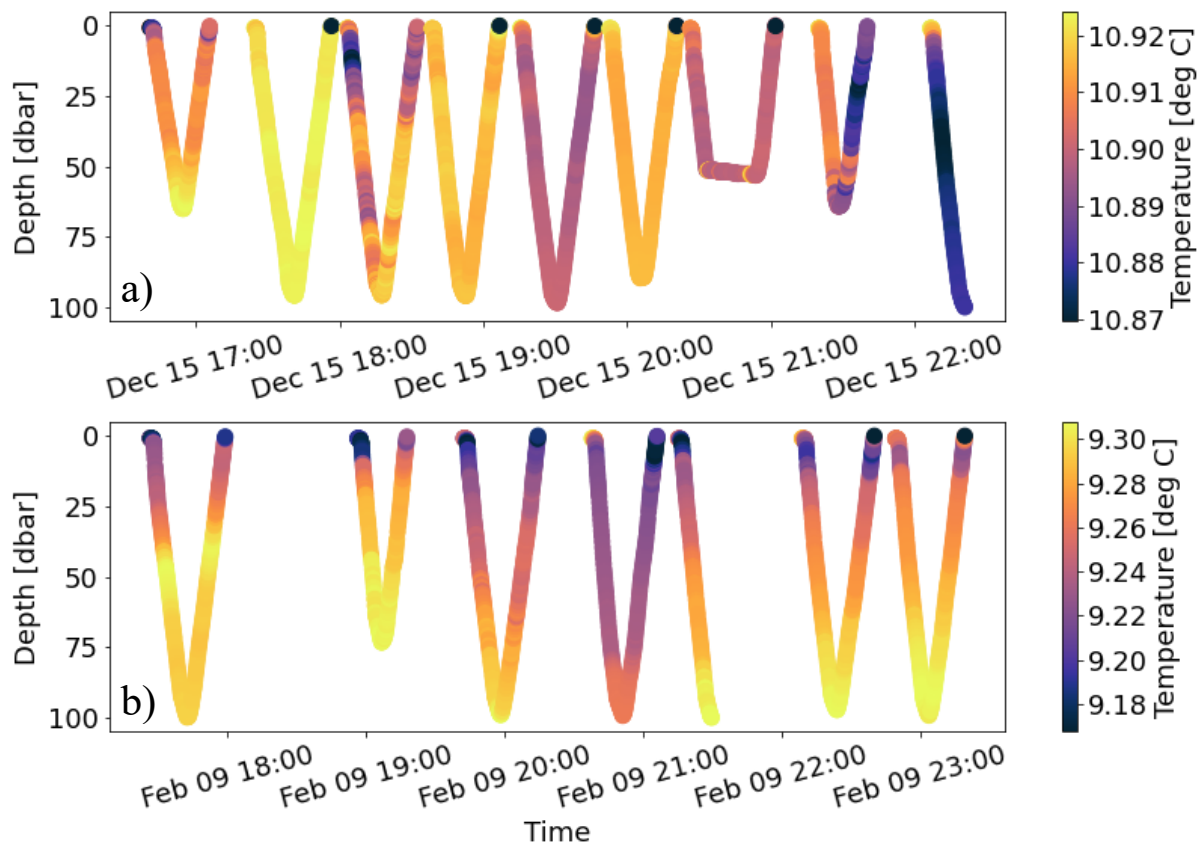
299 9.300 °C and measured salinity varied between 28 and 29.5 PSU (Figure 5b). Density varied by  
 300 about one sigma theta during all dives with surface sigma theta between 21.8 and 22.6 and sigma  
 301 theta at depth between 22.6 and 22.8 (Figure 5b). Seaglider dive depth did not exceed 100 m  
 302 during both deployments due to local bathymetry. The density structure of Colvos Passage  
 303 (Figure 6a) is well-mixed with density varying by less than 0.06 kg m<sup>-3</sup> over 100 m. Some dives  
 304 during this deployment experienced unstable water columns with less dense water sitting below  
 305 dense water. The western side of Blake Island (Figure 6b) was well-stratified with sigma theta



**Figure 6.** a) The dive path of SG175 during the 15 December 2023 deployment in Colvos Passage plotted as depth v.s. time with color indicating water density along the glider’s path. b) The dive path of SG175 during the 09 February 2024 deployment near Blake Island plotted as depth v.s. time with color indicating water density along the glider’s path.

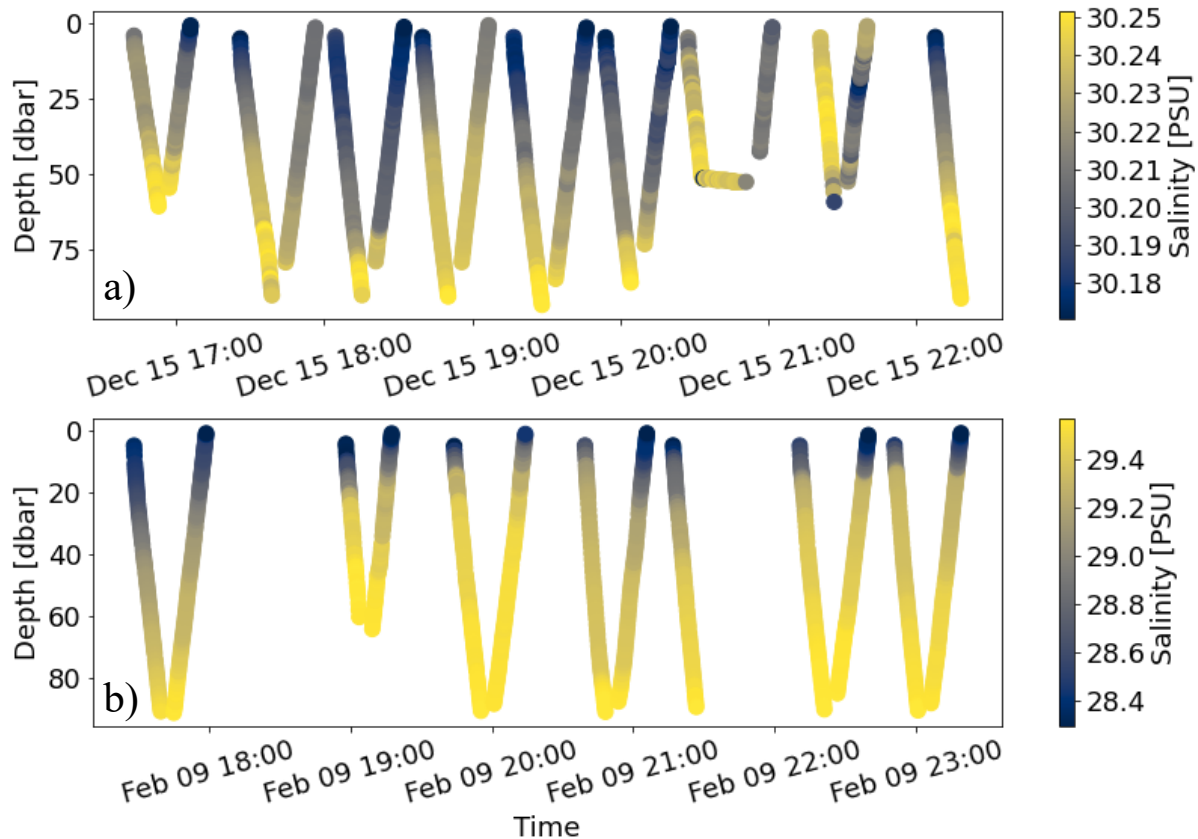
306 varying by 0.8 kg m<sup>-3</sup> over 100 m. The Seaglider experienced stable water columns for all dives  
 307 during this deployment.

308 Water temperature during both dives varied by less than 0.5 °C (Figure 7). The 15  
 309 December 2023 deployment saw water temperatures between 10.87 and 10.92 °C (Figure 7a);  
 310 the water column was well-mixed and does not exhibit a clear depth-dependent signal based on  
 311 temperature. Southern Colvos Passage (left-hand side of Figure 7a) was warmer than the  
 312 northern end (right-hand side of Figure 7a) where Colvos Passage meets the main basin of Puget  
 313 Sound. The 09 February 2024 deployment of SG175 found water temperatures between 9.18 and  
 314 9.30 °C (Figure 7b). The water column was stratified with temperature increasing with depth  
 315 (Figure 7b). Each dive during this deployment exhibited similar depth-dependent temperature  
 316 signals.



**Figure 7.** a) The dive path of SG175 during the 15 December 2023 deployment in Colvos Passage plotted as depth v.s. time with color indicating water temperature (°C) along the glider’s path. b) The dive path of SG175 during the 09 February 2024 deployment near Blake Island plotted as depth v.s. time with color indicating water temperature (°C) along the glider’s path.

317 Salinity in Colvos Passage varied by no more than about one PSU during both  
 318 deployments (Figure 8). The first deployment found salinities between 30.17 and 30.25 PSU  
 319 (Figure 8a). Salinity generally increased with depth. Surface salinity was between 30.18 and  
 320 ~30.22 PSU and salinity at depth was consistently between 30.24 and 30.25 PSU (Figure 8a).  
 321 The second deployment found salinities between 28.3 and 29.5 PSU (Figure 8b). Surface



**Figure 8.** a) The dive path of SG175 during the 15 December 2023 deployment in Colvos Passage plotted as depth v.s. time with color indicating water salinity (PSU) along the glider's path. b) The dive path of SG175 during the 09 February 2024 deployment near Blake Island plotted as depth v.s. time with color indicating water salinity (PSU) along the glider's path.

322 salinities varied between 28.3 and 28.8 PSU and salinity at depth ranged between 29.2 and 29.5

323 PSU. Salinity increased with depth, causing a stratified water column.

324 Depth-averaged salinity measured by the Seaglider's conductivity cell during both

325 deployments was 30.068 PSU with a standard error of 0.093. The average calculated salinity

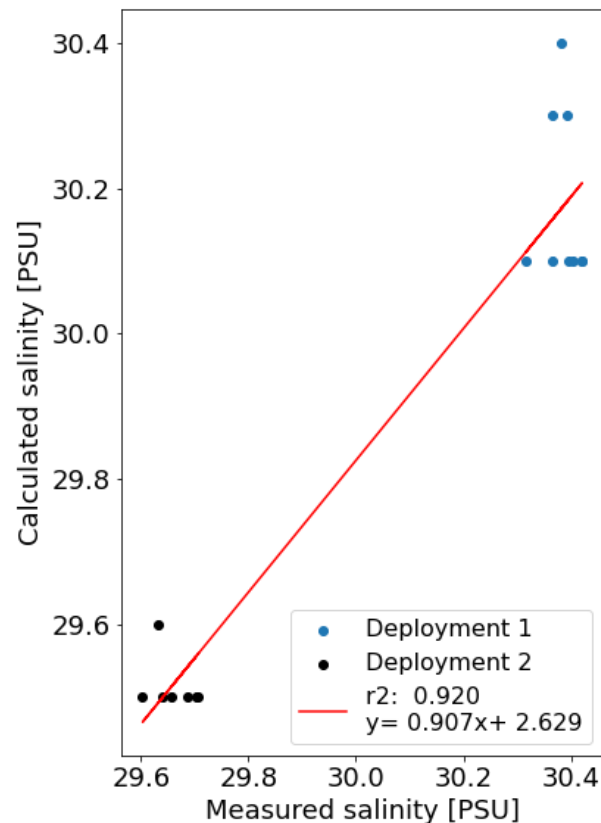
326 using the Seaglider's density at neutral buoyancy, in situ temperature, and pressure was 29.89

327 PSU with a standard error of 0.088. A T-test of the data showed the means of the two salinity  
 328 measurement methods are not statistically different (t-statistic: 1.412, p-value: 0.168). This  
 329 indicates no significant difference between the measured and calculated salinity values.  
 330 Calculated and measured salinity at apogee from both deployments were linearly regressed and  
 331 yielded a slope of 0.907, an r-squared value  
 332 of 0.92, and a p-value < 0.001 (Figure 9),  
 333 showing that the regression is significant.  
 334 Two distinct water masses were sampled  
 335 during the two deployments with salinity  
 336 estimates from deployment two being more  
 337 tightly grouped than from deployment one.

338

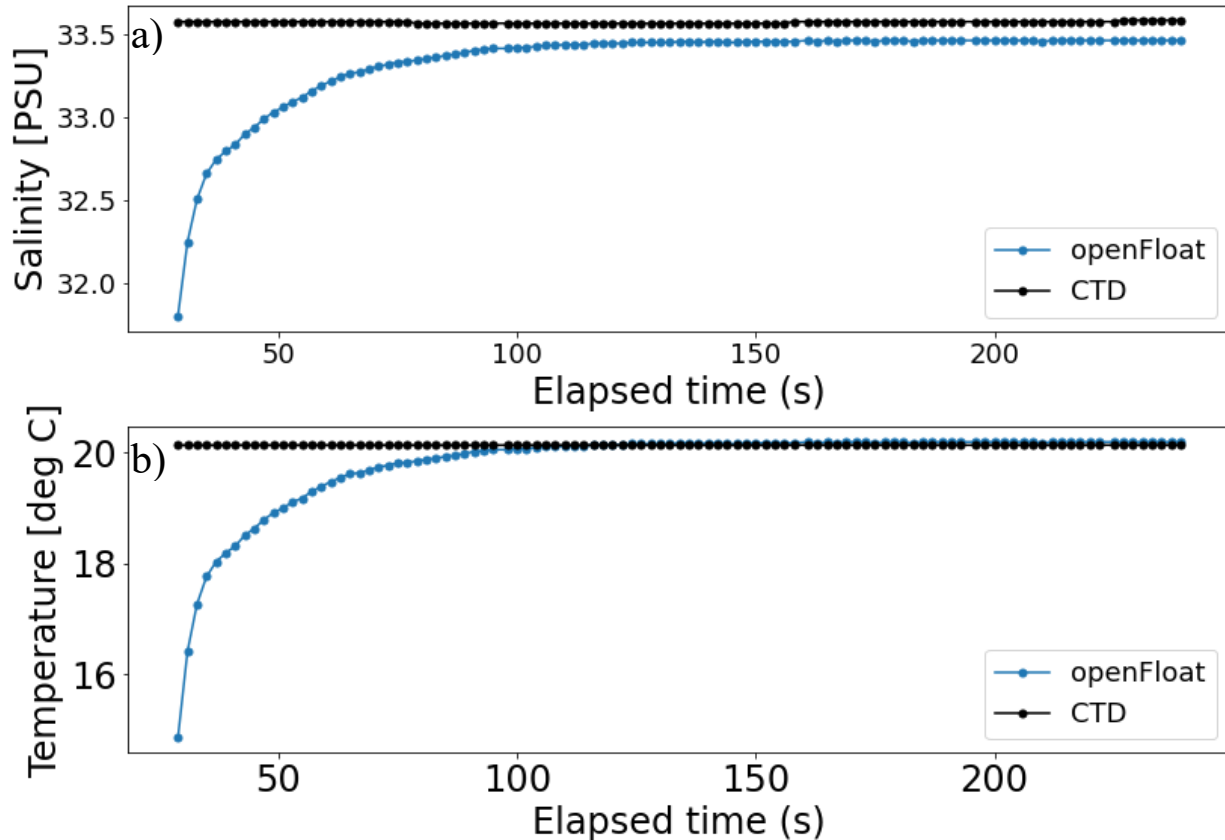
### 339 *openFloat salinity calculations*

340 The openFloat performed similar salinity  
 341 estimations in a four-meter-deep saline test  
 342 tank in the Ocean Sciences Building  
 343 (University of Washington). These  
 344 measurements were compared to a YSI EXO  
 345 Sonde CTD (Figure 10a). The YSI CTD  
 346 measured tank salinity to be  $33.575 \pm 0.006$   
 347 PSU (mean  $\pm$  SD, Figure 10a) and tank temperature as  $20.137 \pm 0.001$  °C (mean  $\pm$  SD, Figure  
 348 10b). Salinity determined from the openFloat was  $33.46 \pm 0.01$  PSU (mean  $\pm$  SD, Figure 10a)  
 349 upon the equilibration of its temperature sensor to  $20.17 \pm 0.03$  °C (mean  $\pm$  SD, Figure 10b) after



**Figure 9.** Calculated salinity using Seaglider density at neutral buoyancy during apogee plotted against the expected salinity during apogee overlaid with a linear regression. Expected salinity is calculated as the mean of salinity at the beginning and end of apogee.

350 ~100 seconds. The openFloat's salinity estimations were  $0.11 \pm 0.01$  PSU (mean  $\pm$  SD, Figure  
 351 10a) less than the salinity measured by the YSI EXO Sonde. The mean difference in temperature  
 352 measured by the openFloat and YSI EXO Sonde after equilibration was  $0.037 \pm 0.028$  °C. This  
 353 analysis also showed the response time of the openFloat's temperature sensor to be  
 354 approximately 100 s.



**Figure 10.** a) Test tank salinity (PSU) measurements made by a YSI EXO Sonde CTD and the openFloat through back-calculation plotted against elapsed time during sampling. b) Temperature (°C) measured by a YSI EXO Sonde CTD and the openFloat plotted against elapsed time during sampling. Temperature measured by the openFloat was inter-calibrated with the YSI to obtain accurate values.

355

## 356 Discussion

### 357 *Summary of results*

358 This study determined that salinity can be estimated from buoyancy-driven AUVs not  
 359 equipped with a traditional conductivity cell-based sensor. Field-based salinity estimations were

360 performed with Seaglider SG175 in Colvos Passage, WA, during two deployments. Using  
361 SG175's in situ measured variables (e.g., temperature and pressure) and known mass and volume  
362 at apogee, salinity was measured to be  $29.9 \pm 0.34$  PSU (mean  $\pm$  SD). The mean difference  
363 between the expected salinity and the calculated salinity was  $0.18 \pm 0.10$  PSU (mean  $\pm$  SD).  
364 Likewise, the openFloat, when neutrally buoyant, estimated test tank salinity using onboard  
365 temperature and pressure readings within  $0.11 \pm 0.01$  PSU (mean  $\pm$  SD) of the true salinity as  
366 determined by a YSI EXO Sonde CTD. These accuracies are well within the range of variability  
367 in coastal Puget Sound (Babson et al., 2006; Moore et al., 2008; MacCready et al., 2021),  
368 indicating this method exhibits promise for use in these types of environments.

369

### 370 *Reasons that contribute to errors in calculations*

371       These salinity calculations are currently limited by several factors. Primarily, they are  
372 most limited by mass and volume measurements which govern calculations of AUV density.  
373 Seagliders are large and weigh roughly 50 kg; typical scales used to weigh a Seaglider are  
374 accurate to one gram. Additionally, obtaining an accurate physical measure of an AUV's volume  
375 can be challenging. Digital designs offer an idealized estimate of what the expected volume is,  
376 but this may not reflect the AUV's true volume after manufacturing. The volume of a buoyancy-  
377 driven AUV can be determined by making the AUV neutrally buoyant in a body of water with  
378 known density (e.g., a test tank). The AUV's volume can be calculated by dividing its known  
379 mass by known density; this density can be obtained with a CTD. While measuring the physical  
380 volume of an AUV is challenging, obtaining the change in volume of an AUV due to changes in  
381 its buoyancy system is known to a high level of accuracy. Both Seagliders and the openFloat use  
382 motor-controlled buoyancy systems capable of sub-100 $\mu$ m positional accuracy.

383 For Seagliders specially, the assumption of neutral buoyancy in the field may not be fully met  
384 since the glider is never in a completely stationary position at one depth, which may influence  
385 the calculation accuracy. Prior to deployment, Seagliders are ballasted and tuned to be nearly  
386 neutrally buoyant at the expected maximum depth. Seagliders are not designed to maintain depth  
387 at neutral buoyancy, but rather momentarily pass through neutral buoyancy as their buoyancy  
388 engines begin to pump during the apogee maneuver. Since Seagliders become neutrally buoyant  
389 once per dive cycle, this results in one salinity point estimation per dive. The assumption of a  
390 stationary AUV at neutral buoyancy is not fully met in the salinity calculation with SG175 since  
391 Seagliders never fully attain neutral buoyancy (e.g., parking at one depth). This likely limits the  
392 accuracy of the salinity estimation using the Seaglider platform.

393 For the openFloat specifically, limitations of off-the-shelf sensors likely contribute to how  
394 well salinity can be calculated. The pressure sensor's temperature readings have a slow response  
395 time to equilibrate with environmental conditions due to the pressure sensor being potted in  
396 epoxy and are limited in resolution based in the sensor's 22-bit ADC. This limits the speed at  
397 which temperature readings can be made and introduces errors if salinity is calculated too  
398 quickly. However, tests in the pool helped constrain the time required (100 s) for the float to park  
399 in specific conditions such that a better measurement can be made once all sensors have  
400 equilibrated to the environment. Because of the smaller observational resolution of the off-the-  
401 shelf sensors, openFloat data may appear to have smaller standard deviations and errors than  
402 industry-grade sensors.

403

404

405

406 *Suggested use cases and benefits of the method*

407 This method presents several potential benefits. Back-calculating salinity this way does not  
408 rely on traditional methods that may be prone to error due to biofouling and sensor drift, or  
409 cracks in the sensors itself. This method of determining salinity can be applied in-post after an  
410 AUV has been deployed for some time since back-calculating salinity does not require a physical  
411 modification to an AUV; it also provides a form of redundancy if a conductivity cell does break.  
412 For existing systems like Argo floats or  $\mu$ floats that park at neutral buoyancy for extended  
413 periods of time, this calculation can be used to create time series data rather than single-point  
414 measurements that Seagliders can obtain during a dive.

415 These salinity calculations are currently limited in depth space, as they can only be  
416 performed when an AUV is neutrally buoyant at the bottom of a profile or is stationary at one  
417 depth. For Seagliders, this occurs briefly during the apogee maneuver as the glider passes from  
418 sinking to floating; profiling floats can maintain neutral buoyancy for extended periods of time,  
419 unlike Seagliders. Because this method cannot currently produce depth profiles, it may be best  
420 suited for a vehicle that either follows a specific water mass at depth or one that performs a duty  
421 cycle to incrementally park floats at different depths as the vehicle ascends to create a depth  
422 profile. This means this method of essentially using an AUV as a density sensor when neutrally  
423 buoyant is most appropriate for profiling floats over Seagliders.

424 Since profiling floats present the best use case for this method, the salinity calculations could  
425 be further tested with global Argo data to expand the salinity range and add more data over  
426 which to test the accuracy of the method. This also presents a way to get salinity data should a  
427 conductivity cell on a float get damaged once deployed and the float becomes inaccessible to  
428 retrieve.

429 For the DIY AUV use case, this method of calculating salinity is best used in regions with  
430 large swings in daily and annual salinity. Off-the-shelf temperature sensors are likely not  
431 accurate enough to be used in locations that experience little variation in salinity. This salinity  
432 estimation method is most appropriate for coastal systems that experience high variability, like  
433 Puget Sound, WA, that exceeds the limitations in accuracy (Babson et al., 2006; Moore et al.,  
434 2008; MacCready et al., 2021). This would not be suitable for the open ocean or deep ocean  
435 where observed changes are smaller than the current accuracy of the method (Lukas and  
436 Santiago-Mandujano, 2008; Stammer et al., 2021).

437

#### 438 *Potential improvements to the openFloat*

439 Future improvements for the openFloat will likely include upgrading the custom circuit  
440 boards to printed circuit boards for ease of mass production. This electrical redesign is likely to  
441 incorporate a different microcontroller as the float's main controller in order to use the one-wire  
442 communication protocol. The ESP32 Feather Huzzah V2 is not currently able to communicate  
443 with sensors such as the ds18b20 temperature probes that are mounted on its sensor cap due to a  
444 known bug in the ESP32 V2 architecture with the one-wire communication protocol. This  
445 temperature sensor is low-cost and has an accuracy of  $\pm 0.1$  °C (Thaler et al., 2024) when  
446 readings are averaged from three sensors (currently equipped on the openFloat). Additionally,  
447 future improvements include performing the openFloat's salinity calculation in a wide range of  
448 salinities and temperatures to quantify the accuracy of its calculations in a variety of  
449 environmental conditions found in Puget Sound. Most improvements to be made are in software  
450 to increase the energy efficiency of the float as a whole.

451

### 452 *Future use of the openFloat*

453 Future tests to validate the openFloat may include side-by-side deployments of the openFloat  
454 alongside a Seaglider for in situ data comparison to a known standard measurement. Colvos  
455 Passage, WA, poses as a unique location for such tests since the water column in nearly  
456 homogenous, ensuring that both AUVs measure similar water parcels. The production and  
457 growth of an openFloat fleet to create a low-cost array of sensors could help increase coastal  
458 observations, document change in real time and help support digital models and forecasting (e.g.,  
459 the Live Ocean Model).

460

### 461 **Conclusions**

462 This project developed a cost-efficient profiling float made from common and easily  
463 producible parts to indirectly measure salinity in dynamic coastal systems to increase the spatial  
464 resolution of salinity measurements. This project successfully validated the indirect  
465 determination of salinity using properties of buoyancy-driven AUVs at neutral buoyancy and  
466 sensor measurements of temperature and pressure. The indirect determination of salinity was  
467 evaluated using both a Seaglider and a DIY profiling float. Calculations from field deployments  
468 of a Seaglider and test tank trials of the openFloat show that salinity can be measured nearly  
469 identically by existing industry-grade and DIY AUVs. This method achieves salinity values to  
470 about  $\pm 0.20$  PSU and is best used in dynamic coastal systems where salinity variation is high.  
471 This establishes a path forward for further development of the openFloat and a salinity  
472 calculation that could be used to extract more data from other buoyancy-driven devices.

473

474

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547

548 **Appendix**

549 All data are made publically available here:

550 [https://github.com/cflaim1123/openFloat\\_thesis/tree/main/data](https://github.com/cflaim1123/openFloat_thesis/tree/main/data).551 **Table 1** openFloat project budget.

Category	Item	#	Price per	Total price
Electrical - general	perfboard (pack of 50)	7	\$0.50	\$3.50
	Misc wires	NA	NA	\$15.00
	Male header pins	NA	NA	\$5.00
	Female header pins	NA	NA	\$5.00
	Solder	NA	NA	\$10.00
	Resistors	NA	NA	\$0.10
	Misc GPS radio wires	NA	NA	\$15.00
Electrical - control	ESP32 feather huzzah v2	2	\$19.95	\$39.90
	BLDC motor	1	\$30.00	\$30.00
	5V, 5A voltage step-down regulator	1	\$32.95	\$32.95
	custom battery pack	1	\$15.00	\$15.00
Electrical - sensing	analog pressure sensor	1	\$20.00	\$20.00
	Thermistor	1	\$2.00	\$2.00
	adafruit data logger feather wing	1	\$8.95	\$8.95
	digital temperature sensor (pack of 5)	1	\$10.00	\$10.00
	BMP180 pressure sensor	1	\$9.95	\$9.95
	DHT22 humidity sensor	1	\$9.95	\$9.95
	TSL2591 lux sensor	1	\$6.95	\$6.95
	Adafruit 9-DOF absolute orientation IMU - BN055	1	\$35.00	\$35.00
	Adafruit ultiamte GPS	2	\$30.00	\$60.00
	915 mHz LoRa radio module	2	\$20.00	\$40.00
	Precision current meter	1	\$10.00	\$10.00
	GPS antenna	1	\$20.00	\$20.00
	Radio antenna	2	\$10.00	\$20.00
Mechanical	Low-pitch, high-precision ACME lead screw	1	\$15.00	\$15.00
	Misc O-ring	NA	NA	\$10.00

	Misc PVC pipe for piston	1	\$10.00	\$10.00
	4" PVC pipe	1	\$20.00	\$20.00
	4" PVC endcaps	2	\$10.00	\$20.00
	metal cable glands	5	\$10.00	\$50.00
	Epoxy	2	\$10.00	\$20.00
			Total:	\$529.25

552