

# Cell abundances of taxonomic groups determined using a custom convolutional neural network from live Imaging FlowCytobot (IFCB) at seven stations sampled during R/V Pt. Sur cruise PS 18-09 in the western Gulf of Mexico, Sept-Oct 2017.

**Website:** <https://www.bco-dmo.org/dataset/840201>

**Data Type:** Cruise Results

**Version:** 1

**Version Date:** 2021-02-08

## Project

» [RAPID: Hurricane Impact on Phytoplankton Community Dynamics and Metabolic Response](#) (HRR)

| Contributors                        | Affiliation   | Role                      |
|-------------------------------------|---|---------------------------|
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## Abstract

Cell abundance data for taxonomic groups from a custom convolutional neural network from live Imaging FlowCytobot (IFCB) at seven stations from surface and chlorophyll maximum depths during R/V Pt. Sur PS 18-09, western Gulf of Mexico, Sept-Oct 2017. This dataset was uncorrected and is for comparison with the Texas Observatory for Algal Succession Time-series (TOAST).

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## Coverage

**Spatial Extent:** N:29.0649 E:-94.9 S:27.2286 W:-97.268

**Temporal Extent:** 2017-09-23 - 2017-10-01

## Methods & Sampling

On each of 2 cruise legs 01 and 03, samples were collected at 7 stations (S01, S06, S11, S16, S21, SS, and GI) from 2 depths [surface and chlorophyll maximum depth when possible; see HRR-bottle data] by CTD-rosette. At each station, triplicate 5-ml samples pre-filtered through 150  $\mu$ m Nitex were analyzed immediately with an onboard Imaging FlowCytobot. All image data can be viewed on the TOAST dashboard: [https://toast.tamu.edu/timeline?dataset=HRR\\_cruise](https://toast.tamu.edu/timeline?dataset=HRR_cruise).

Image analysis and feature extraction were performed using software developed by Sosik and colleagues which is available on github (<https://github.com/hsosik/ifcb-analysis/>). The automated classification approach of

Sosik & Olson (2007), as modified and described by Anglès et al. (2019), was employed and the automated classification results were then inspected visually and **manually** corrected into a total of 102 categories that included 35 categories of diatoms, 30 categories of dinoflagellates, 10 categories of ciliates, 10 categories of flagellates, and 17 'others', which included filamentous cyanobacteria, freshwater chlorophytes, coccolithophorids, and small cells that could not be identified taxonomically from images (refer to Fiorendino et al. 2021. for more details).

For comparison with the Texas Observatory for Algal Succession Time-series (TOAST), IFCB images were also classified automatically into one of 112 classes utilizing a custom convolutional neural network (**CNN**) trained on a curated set of images (Henrichs et al. 2021.). See related dataset.

**Biomass** for each image was estimated using the algorithm developed by Moberg & Sosik (2012) to calculate cellular volume from the extracted image features and then convert to total carbon per image (Menden-Deuer & Lessard 2000) and summed for each class. See related dataset.

Sampling locations:

| Sample ID | Station | Leg | Location         |
|-----------|---------|-----|------------------|
|           |         |     | Lat °N/Long °W   |
| L1_S01    | S01     | 1   | 27.2286 -97.2686 |
| L3_S01    | S01     | 3   |                  |
| L1_S06    | S06     | 1   | 27.8358 -96.9874 |
| L3_S06    | S06     | 3   |                  |
| L1_S11    | S11     | 1   | 28.2614 -96.4129 |
| L3_S11    | S11     | 3   |                  |
| L1_S16    | S16     | 1   | 28.5366 -95.8656 |
| L3_S16    | S16     | 3   |                  |
| L1_S21    | S21     | 1   | 28.7644 -95.2978 |
| L3_S21    | S21     | 3   |                  |
| L1_SS     | SS      | 1   | 28.9600 -95.0946 |
| L3_SS     | SS      | 3   |                  |
| L1_GI     | GI      | 1   | 29.0649 -94.9000 |
| L3_GI     | GI      | 3   |                  |

## Data Processing Description

### - BCO-DMO Processing Notes:

- data submitted in Excel file "manual\_carbon\_ifcb.xlsx" sheet "Biomass" extracted to csv
- added conventional header with dataset name, PI name, version date
- modified parameter names to conform with BCO-DMO naming conventions
- rounded values to 2 decimal places

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## Data Files

| File  |
|---|
| <b>IFCB_cnn_count.csv</b> (Comma Separated Values (.csv), 15.00 KB)<br>MD5:bcc6bc821dabdc2a1e5ada6e5d65179f |
| Primary data file for dataset ID 840201   |

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## Related Publications

Anglès, S., Jordi, A., Henrichs, D. W., & Campbell, L. (2019). Influence of coastal upwelling and river discharge on the phytoplankton community composition in the northwestern Gulf of Mexico. *Progress in Oceanography*, 173, 26–36. doi:[10.1016/j.pocean.2019.02.001](https://doi.org/10.1016/j.pocean.2019.02.001)

*Results*

Fiorendino, J. M., Gaonkar, C. C., Henrichs, D. W., & Campbell, L. (2021). Drivers of microplankton community assemblage following tropical cyclones. *Journal of Plankton Research*, 45(1), 205–220.

<https://doi.org/10.1093/plankt/fbab073>

*Results*

Henrichs, D. W., Anglès, S., Gaonkar, C. C., & Campbell, L. (2021). Application of a convolutional neural network to improve automated early warning of harmful algal blooms. *Environmental science and pollution research international*, 28(22), 28544–28555. <https://doi.org/10.1007/s11356-021-12471-2>

*Results*

Menden-Deuer, S., & Lessard, E. J. (2000). Carbon to volume relationships for dinoflagellates, diatoms, and other protist plankton. *Limnology and Oceanography*, 45(3), 569–579. doi:[10.4319/lo.2000.45.3.0569](https://doi.org/10.4319/lo.2000.45.3.0569)

*Methods*

Moberg, E. A., & Sosik, H. M. (2012). Distance maps to estimate cell volume from two-dimensional plankton images. *Limnology and Oceanography: Methods*, 10(4), 278–288. doi:[10.4319/lom.2012.10.278](https://doi.org/10.4319/lom.2012.10.278)

*Methods*

Olson, R. J., & Sosik, H. M. (2007). A submersible imaging-in-flow instrument to analyze nano- and microplankton: Imaging FlowCytobot. *Limnology and Oceanography: Methods*, 5(6), 195–203.

doi:[10.4319/lom.2007.5.195](https://doi.org/10.4319/lom.2007.5.195)

*Methods*

Sosik, H. M., & Olson, R. J. (2007). Automated taxonomic classification of phytoplankton sampled with imaging-in-flow cytometry. *Limnology and Oceanography: Methods*, 5(6), 204–216. doi:[10.4319/lom.2007.5.204](https://doi.org/10.4319/lom.2007.5.204)

*Methods*

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## Related Datasets

### IsRelatedTo

Campbell, L., Henrichs, D. W. (2021) **Cell abundances for taxonomic groups from manually corrected**

**live Imaging FlowCytobot (IFCB) analysis of water samples collected from surface and chlorophyll maximum depths during R/V Pt. Sur cruise PS 18-09 in the western Gulf of Mexico, Sept-Oct 2017.** Biological and Chemical Oceanography Data Management Office (BCO-DMO). (Version 1) Version Date 2021-02-08 doi:10.26008/1912/bco-dmo.840060.1 [[view at BCO-DMO](#)]

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## Parameters

| Parameter      | Description  | Units            |
|----------------|--|------------------|
| Major_Category | 5 major groups of microplankton  | unitless         |
| Class          | Taxonomic names of the species or genus (if known) or group identified | unitless         |
| L1GI_1         | Cell abundance of class at Leg 1 station GI; replicate 1               | cells/milliliter |
| L1GI_2         | Cell abundance of class at Leg 1 station GI; replicate 2               | cells/milliliter |
| L1GI_3         | Cell abundance of class at Leg 1 station GI; replicate 3               | cells/milliliter |
| L1SS_1         | Cell abundance of class at Leg 1 station SS; replicate 1               | cells/milliliter |
| L1SS_2         | Cell abundance of class at Leg 1 station SS; replicate 2               | cells/milliliter |
| L1SS_3         | Cell abundance of class at Leg 1 station SS; replicate 3               | cells/milliliter |
| L1S21_1        | Cell abundance of class at Leg 1 station S21; replicate 1              | cells/milliliter |
| L1S21_2        | Cell abundance of class at Leg 1 station S21; replicate 2              | cells/milliliter |
| L1S21_3        | Cell abundance of class at Leg 1 station S21; replicate 3              | cells/milliliter |
| L1S16_1        | Cell abundance of class at Leg 1 station S16; replicate 1              | cells/milliliter |
| L1S16_2        | Cell abundance of class at Leg 1 station S16; replicate 2              | cells/milliliter |
| L1S16_3        | Cell abundance of class at Leg 1 station S16; replicate 3              | cells/milliliter |
| L1S11_1        | Cell abundance of class at Leg 1 station S11; replicate 1              | cells/milliliter |
| L1S11_2        | Cell abundance of class at Leg 1 station S11; replicate 2              | cells/milliliter |
| L1S11_3        | Cell abundance of class at Leg 1 station S11; replicate 3              | cells/milliliter |
| L1S06_1        | Cell abundance of class at Leg 1 station S06; replicate 1              | cells/milliliter |
| L1S06_2        | Cell abundance of class at Leg 1 station S06; replicate 2              | cells/milliliter |
| L1S06_3        | Cell abundance of class at Leg 1 station S06; replicate 3              | cells/milliliter |
| L1S01_1        | Cell abundance of class at Leg 1 station S01; replicate 1              | cells/milliliter |
| L1S01_2        | Cell abundance of class at Leg 1 station S01; replicate 2              | cells/milliliter |
| L1S01_3        | Cell abundance of class at Leg 1 station S01; replicate 3              | cells/milliliter |
| L3GI_1         | Cell abundance of class at Leg 3 station GI; replicate 1               | cells/milliliter |
| L3GI_2         | Cell abundance of class at Leg 3 station GI; replicate 2               | cells/milliliter |
| L3GI_3         | Cell abundance of class at Leg 3 station GI; replicate 3               | cells/milliliter |
| L3SS_1         | Cell abundance of class at Leg 3 station SS; replicate 1               | cells/milliliter |
| L3SS_2         | Cell abundance of class at Leg 3 station SS; replicate 2               | cells/milliliter |
| L3SS_3         | Cell abundance of class at Leg 3 station SS; replicate 3               | cells/milliliter |
| L3S21_1        | Cell abundance of class at Leg 3 station S21; replicate 1              | cells/milliliter |
| L3S21_2        | Cell abundance of class at Leg 3 station S21; replicate 2              | cells/milliliter |
| L3S21_3        | Cell abundance of class at Leg 3 station S21; replicate 3              | cells/milliliter |
| L3S16_1        | Cell abundance of class at Leg 3 station S16; replicate 1              | cells/milliliter |

|         |   |                  |
|---------|---|------------------|
| L3S16_2 | Cell abundance of class at Leg 3 station S16; replicate 2 | cells/milliliter |
| L3S16_3 | Cell abundance of class at Leg 3 station S16; replicate 3 | cells/milliliter |
| L3S11_1 | Cell abundance of class at Leg 3 station S11; replicate 1 | cells/milliliter |
| L3S11_2 | Cell abundance of class at Leg 3 station S11; replicate 2 | cells/milliliter |
| L3S11_3 | Cell abundance of class at Leg 3 station S11; replicate 3 | cells/milliliter |
| L3S06_1 | Cell abundance of class at Leg 3 station S06; replicate 1 | cells/milliliter |
| L3S06_2 | Cell abundance of class at Leg 3 station S06; replicate 2 | cells/milliliter |
| L3S06_3 | Cell abundance of class at Leg 3 station S06; replicate 3 | cells/milliliter |
| L3S01   | Cell abundance of class at Leg 3 station S01              | cells/milliliter |

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## Instruments

|   |  |
|---|--|
| <b>Dataset-specific Instrument Name</b> | Image FlowCytobot (McLane Research Laboratories, Inc.)   |
| <b>Generic Instrument Name</b>          | Imaging FlowCytobot  |
| <b>Generic Instrument Description</b>   | The Imaging FlowCytobot (IFCB) is an in-situ automated submersible imaging flow cytometer that generates images of particles in-flow taken from the aquatic environment. <a href="https://mclanelabs.com/imaging-flowcytobot/">https://mclanelabs.com/imaging-flowcytobot/</a> |

|   |   |
|---|---|
| <b>Dataset-specific Instrument Name</b> |   |
| <b>Generic Instrument Name</b>          | Niskin bottle   |
| <b>Dataset-specific Description</b>     | Used to collect samples   |
| <b>Generic Instrument Description</b>   | A Niskin bottle (a next generation water sampler based on the Nansen bottle) is a cylindrical, non-metallic water collection device with stoppers at both ends. The bottles can be attached individually on a hydrowire or deployed in 12, 24, or 36 bottle Rosette systems mounted on a frame and combined with a CTD. Niskin bottles are used to collect discrete water samples for a range of measurements including pigments, nutrients, plankton, etc. |

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## Deployments

PS1809

|                    |  |
|--------------------|--|
| <b>Website</b>     | <a href="https://www.bco-dmo.org/deployment/784313">https://www.bco-dmo.org/deployment/784313</a>  |
| <b>Platform</b>    | R/V Point Sur  |
| <b>Start Date</b>  | 2017-09-22   |
| <b>End Date</b>    | 2017-10-03   |
| <b>Description</b> | HRR study with three legs. Chief Scientists: Steve DiMarco (Leg 1); Kristen Thyng (Leg 2); Lisa Campbell (Leg 3). R2R Cruise Page: <a href="https://www.rvdata.us/search/cruise/PS1809">https://www.rvdata.us/search/cruise/PS1809</a> |

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## Project Information

### **RAPID: Hurricane Impact on Phytoplankton Community Dynamics and Metabolic Response (HRR)**

**Coverage:** Texas coast

#### *NSF Award Abstract:*

Hurricane Harvey is the strongest hurricane to hit the Texas coast in decades and the resulting tidal surges, flooding and terrestrial runoff have had a severe impact on the coastal ocean. The effects on the phytoplankton, the first link in the food chain, may be unprecedented. To determine how the phytoplankton community will respond to such drastic changes in salinity, nutrient inputs, and potential toxins, immediate and continuous sampling is the only way to fully capture the effects and to identify when conditions return to "normal". An automated, continuous phytoplankton imaging instrument that is deployed on the Texas coast records images of the phytoplankton and permits calculation of the abundance of different species. Together with molecular information on the genes that have been "turned on", or expressed, outcomes of this project will help determine the responses of individual types of phytoplankton. Extreme storms are expected to increase in frequency with future climate change, so the responses identified now will be valuable in predicting how such events will affect these primary producers, which in turn support most of the food webs in marine ecosystems, in the future.

High temporal resolution observations from the Imaging FlowCytobot (IFCB) have revealed that hurricanes in the Gulf of Mexico cause drastic changes in the phytoplankton community structure. The objectives of this RAPID project are: 1) to characterize the dynamics of the phytoplankton species in relation to the environmental variables along the Texas coast; 2) to assess the short and long-term changes in the phytoplankton community; and 3) to identify the strategies of the phytoplankton community for resource acquisition. To accomplish these objectives, this project will utilize IFCB time series to follow phytoplankton community structure during the recovery period from Hurricane Harvey. In addition, two RAPID response cruises (in late September and early October) to sample at 5 sites along a transect from Galveston to Port Aransas, TX. At each station, CTD profiles and water samples from surface and the chlorophyll maximum will be collected for nutrients, carbonate chemistry, and RNA sequencing for metatranscriptomic analysis. Metatranscriptomics can provide an indication of the metabolic strategies employed and functional relationships within the plankton community in response to changes in the environment. The advantage of a metatranscriptomic approach is that the entire molecular response to the environment is captured. So, while the response of phytoplankton to increased nutrient inputs from floodwater runoff is targeted, the responses to other environmental stresses (toxics, hypoxia, acidification) are also captured. Analyses of this time series using multivariate statistical techniques, such as principal component analysis (PCA), and network analysis, a powerful technique for identifying potential interactions among taxa, will provide insights on the environmental factors and metabolic responses structuring the community during the aftermath of the hurricane.

**Related data from the The Texas Observatory for Algal Succession Time-Series (TOAST) can be found at the following:** [https://toast.tamu.edu/timeline?dataset=HRR\\_Cruise](https://toast.tamu.edu/timeline?dataset=HRR_Cruise)

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## Funding

| Funding Source   | Award                       |
|--|-----------------------------|
| <a href="#">NSF Division of Ocean Sciences (NSF OCE)</a> | <a href="#">OCE-1760620</a> |

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