

Decadal Visioning Workshop 2024 THE FUTURE OF COASTAL

PROCESSES RESEARCH



JUNE 11 – JUNE 13

Hilton St. Petersburg Bayfront – 333 1st Street, SE

Decadal Visioning Workshop THE FUTURE OF COASTAL PROCESSES RESEARCH

Examples of Applied Research Opportunities to Implement Solutions



Examples of Applied Research Opportunities to Implement Solutions

Session Objective: Give examples of applied research to spur thinking in breakouts.

PRESENTERS

Meg Palmsten, Research Oceanographer, U.S. Geological Survey

Greg Dusek, Senior Scientist, National Oceanographic and Atmospheric Administration

Kate Brodie, Senior Research Oceanographer, U.S. Army Corps of Engineers

Decadal Visioning Workshop: THE FUTURE OF COASTAL PROCESSES RESEARCH



Examples of Applied Research Opportunities to Implement Solutions

Meg Palmsten

Research Oceanographer, U.S. Geological Survey

Decadal Visioning Workshop: THE FUTURE OF COASTAL PROCESSES RESEARCH

Photo credit: Donny Bowers

Operational modeling for decision making

Meg Palmsten, USGS Spicer Bak, U.S. Army Corps of Engineers Tracy Fanara NOAA/National Ocean Service Trey Flowers, NOAA/NWS/Office of Water Prediction Ty Hesser, U.S. Army Corps of Engineers Isabel Houghton, Sofar Ocean Philip Orton, Stevens Institute Allison Penko, U.S. Naval Research Laboratory Kristen Splinter, University of New South Wales Ap Van Dongeren, Deltares John Warner, USGS

This information is preliminary and is subject to revision. It is being provided to meet the need for timely best science. The information is provided on the condition that neither the U.S. Geological Survey nor the U.S. Government shall be held liable for any damages resulting from the authorized or unauthorized use of the information.

Defining operational modeling

- Fast models at high resolution and short timescales
- Wave propagation and compound flooding
- Morphologic evolution and sediment transport
- Urban environment and human infrastructure
- Human and ecosystem health and safety
- Data assimilation
- Uncertainty
- Communication with stakeholders

"....requires development of real time predictive models..." Elko et al. 2015

Significant advancements: technology





Computational power – cloud and operational high-performance computing



Improved data and model data access



Meta-models and Machine Learning

Better boundary conditions and longer lead times

Significant advancements: products

Compound Flooding (National Water Model)



Rip Currents (NOAA/Nearshore Wave Prediction System)



Dune erosion/Total Water Level and Coastal Change (USGS/Nearshore Wave Prediction System)



Impact Modeling (NOPP Hurricane Coastal Impacts Project/Deltares)



Probabilistic information (USACE Coastal Hazards Rapid Prediction System)







Significant advancements: impacts



- More information
- More time to react
- Actionable information by including uncertainties

After Hurricane Idalia, Gulfport, Florida Photo credit: Erica Harris



Technical challenges

01 Cor Rur Imp

Computational limitations Runtime limitations Improving model efficiency

02

Data access Model data dissemination Data and quality control in real time

03

Cybersecurity



Non-technical challenges and opportunities: research community



Non-technical challenges and opportunities: communication



- Explain forecast production
 Consider product design
 Convey uncertainty
 Consider language barriers
- Use relevant variables
- Engage with experienced science communicators
- Run mock events

Promising areas



- Data assimilation
- Multi-fidelity models
- Coupled systems
- Observations and data
- Model optimization and computing
- Machine learning
- Communications

What's missing from nearshore operational models?



- What are the gaps?
- How do we co-develop useful products?
- What are the limitations? (technical and non-technical)



Greg Dusek

Senior Scientist, National Oceanographic and Atmospheric Administration

Decadal Visioning Workshop: THE FUTURE OF COASTAL PROCESSES RESEARCH

Coastal Applications of Artificial Intelligence

Greg Dusek Senior Scientist NOAA National Ocean Service

Applied Research Opportunities to Implement Solutions 2024 USCRP Decadal Visioning Workshop June 12, 2024



Number of Machine Learning + Coastal publications per year



Data from Web of Science

November 30th, 2022 Launch of ChatGPT











Can you provide one sentence which describes how AI might transform coastal ocean research over the next decade?

Al is poised to revolutionize coastal ocean research over the next decade by enabling more accurate and efficient data collection, enhancing predictive models for climate and environmental changes, and facilitating real-time monitoring and management of marine ecosystems.

November 30th, 2022 Launch of ChatGPT











What about if you were J.R.R. Tolkien?

In the coming decade, AI, much like a wizard wielding newfound power, shall transform the realm of coastal ocean research, bestowing upon us the ability to gather data with unparalleled precision, forge predictive models as reliable as the ancient tomes, and oversee the stewardship of marine ecosystems with vigilance akin to the watchful eyes of the Great Eagles.

Outline

Recent Federal Government focus on Al

How has AI been applied to coastal sciences?

- Computer vision
- Prediction and forecasting
- Data analysis

AI Challenges and Limitations

Where are we headed with AI over the next decade?

What is artificial intelligence?

The ability for computational systems to perform tasks that normally require human intelligence

Artificial Intelligence Methods for computer systems to perform human tasks Machine Learning **Expert Systems** Mathematical models with specified structure **Statistics** Operate autonomously learn to perform tasks from data **Foundational Techniques** with human specified and Training Principles rules. (e.g. fuzzy logic) **Deep Learning** Neural networks with multiple specialized layers for encoding structural information

Haupt et al. (2022) The History and Practice of AI in the Environmental Sciences. Bull. Amer. Meteor. Soc., 103, E1351–E1370, https://doi.org/10.1175/BAMS-D-20-0234.1.



OCTOBER 30, 2023

Executive Order on the Safe, Secure, and Trustworthy Development and Use of Artificial Intelligence

- BRIEFING ROOM + PRESIDENTIAL ACTIONS
- National AI Advisory Committee
- Federal Gov AI use case inventory
- 2023 National AI R&D Strategic Plan
- OMB Policy to Advance Governance, Innovation and Risk Management in Federal Agencies' Use of Al



NDAA Artificial Intelligence

Strategic Plan 2021–2025









Welcome to the 5th NOAA Workshop on Leveraging AI in Environmental Sciences

How has AI been applied to coastal sciences?

Some great review articles!

Haupt et al. (2022) The History and Practice of Al in the Environmental Sciences. Bull. Amer. Meteor. Soc., 103, E1351–E1370, <u>https://doi.org/10.1175/BAMS-D-20-0234.1</u>

Goldstein EB et al (2019) A review of machine learning applications to coastal sediment transport and morphodynamics. EarthScience Reviews 194,97–108.

Vistousek et al. (2023) The future of coastal monitoring through satellite remote sensing. Cambridge Prisms: Coastal Futures, 1, e10, 1–18 <u>https://doi.org/10.1017/cft.2022.4</u>

Rubbens et al. (2023) Machine learning in marine ecology: an overview of techniques and applications. ICES Journal of Marine Science Volume 80, Issue 7, Pages 1829–1853, <u>https://doi.org/10.1093/icesjms/fsad100</u>

Computer Vision Detection

Marine Debris Detection



Parrish, C.E., et al. (2023) Uncrewed Aircraft Systems, Machine Learning, and Polarimetric Imaging for Enhanced Marine Debris Shoreline Surveys. Final Report. NOAA Technical Memorandum NOS NCCOS 312. Silver Spring, MD. 31 pp. doi: 10.25923/337h-k518

Wet/Dry Shoreline Detection



Vicens-Miquel M et al (2022) A Deep Learning Based Method to Delineate the Wet/Dry Shoreline and Compute Its Elevation Using High-Resolution UAS Imagery. *Remote Sensing*. 2022; 14(23):5990. https://doi.org/10.3390/rs14235990

HAB Detection



Hardison et al. (2019) HABscope: A tool for use by citizen scientists to facilitate early warning of respiratory irritation caused by toxic blooms of Karenia brevis. PLoS ONE 14(6): e0218489. https://doi.org/10.1371/journal.pone.0218489

Computer Vision Segmentation and classification

Flood segmentation in webcam imagery



Wang et al. (2023) Urban flood extent segmentation and evaluation from realworld surveillance camera images using deep convolutional neural network. Environ Model Softw 105939. https://doi.org/10.1016/j.envsoft.2023.105939

Segmentation of earth surface imagery





Buscombe et al. (2022). Human-in-the-loop segmentation of Earth surface imagery. Earth and Space Science, 9, e2021EA002085. https://doi.org/10.1029/2021EA002085

Point classification of benthic features



I. D. Williams et al. (2019). Leveraging Automated Image Analysis Tools to Transform Our Capacity to Assess Status and Trends of Coral Reefs. Frontiers in Marine Science.

Computer Vision Regression

Optical wave gauging



Buscombe et al. (2020). Optical wave gauging using deep neural networks. Coast Eng. 155, 103593. https://doi.org/10.1016/j.coastaleng.2019.103593. URL: http://www.sciencedirect.com/science/article/pii/S0378383919301243.

Surf zone bathymetry estimates from imagery



Collins et al. (2021). Development of a Fully Convolutional Neural Network to Derive Surf-Zone Bathymetry from Close-Range Imagery of Waves in Duck, NC. Remote Sensing. 13(23):4907. https://doi.org/10.3390/rs13234907

Prediction and Forecasting Data driven models

Predicting particle settling velocity



Goldstein, E. B., and Coco, G. (2014). A machine learning approach for the prediction of settling velocity. Water Resour. Res. 50, 3595–3601. doi: 10.1002/2013WR015116

Prediction of non-tidal residual



Bruneau N. et al. (2020) Estimation of global coastal sea level extremes using neural networks. IOP Environ. Res. Lett 15, 074030. https://doi.org/10.1088/1748-9326/ab89d6

Prediction and Forecasting Surrogate models

ML surrogate for overland urban flooding



Zahura et al. (2020). Training machine learning surrogate models from a high-fidelity physics-based model: Application for real-time street-scale flood prediction in an urban coastal community. Water Resources Research, 56, e2019WR027038. https://doi.org/10.1029/2019WR027038

ML surrogate for wave model nowcasts



Chen et al. (2021) Using machine learning to derive spatial wave data: A case study for a marine energy site. Environ. Modell. Software, 142, 105066, https://doi.org/10.1016/j.envsoft.2021.105066.

Prediction and Forecasting Hybrid models

Predicting future coastal flood risk with a hybrid model



Anderson et al. (2021). Projecting climate dependent coastal flood risk with a hybrid statistical dynamical model. Earth's Future, 9(12), e2021EF002285. https://doi.org/10.1029/2021EF002285

Improving numerical model ensemble averaging with ML



Campos et al. (2020). Improving NCEP's global-scale wave ensemble averages using neural networks. Ocean Model. 101617

Data analysis | Feature extraction

Detecting and extracting sound sources from acoustic data

Automated data extraction/boundary determination **Reflectance Timestack** (c) Prediction Timestack Water Depth Timestack icm Contou each-Water Interface 10-1 0.4 176 176 211 211 211 246 246 246 15 25 15 25 35 45 15 25 Cross-shore distance (m) Cross-shore distance (m) Cross-shore distance (m)

Collins et al., (2023) Automated Extraction of a Depth-Defined Wave Runup Time Series From Lidar Data Using Deep Learning, in IEEE Transactions on Geoscience and Remote Sensing, vol. 61, pp. 1-13, 2023, Art no. 5700913, doi: 10.1109/TGRS.2023.3244488.



White et al. (2022) More than a whistle: automated detection of marine sound sources with a convolutional neural network. Front. Mar. Sci. 9.https://doi.org/10.3389/fmars.2022.879145

Data analysis Recognizing patterns and anomalies

Data clustering and pattern recognition for flood thresholds



Mahmoudi, et al. (2024) Establishing flood thresholds for sea level rise impact communication. *Nat Commun* 15, 4251 (2024). https://doi.org/10.1038/s41467-024-48545-1

Anomaly detection in optical flow fields



de Silva et al. (2023) RipViz: Finding Rip Currents by Learning Pathline Behavior, in *IEEE Transactions on Visualization and Computer Graphics*, doi: 10.1109/TVCG.2023.3243834.

Data analysis Data generation

Super-resolution methods for remotesensed imagery



Wang et al. (2022). A comprehensive review on deep learning based remote sensing image super-resolution methods. Earth-Sci. Rev. 232, 104110. http://dx.doi.org/10.1016/j.earscirev.2022.104110.

Image generation for flood visualization



Lütjens et al. (2023) Physically-consistent generative adversarial networks for coastal flood visualization. arXiv preprint arXiv:2104.04785v4.

Challenges and **Limitations**

Limits of data driven models

Data cleaning and labeling

Explainability

Compute and data storage

Al expertise and workforce development

Where are we headed with AI?

As a dynamic model replacement

nature

NEWS 14 November 2023

DeepMind AI accurately forecasts weather – on a desktop computer

The machine-learning model takes less than a minute to predict future weather worldwide more precisely than other approaches.

By <u>Carissa Wong</u>



Microsoft Research Blog

Introducing Aurora: The first large-scale foundation model of the atmosphere

Published June 3, 2024

By <u>Wessel Bruinsma</u>, Senior Researcher; <u>Megan Stanley</u>, Senior Researcher; <u>Ana Lucic</u>, Researcher; <u>Richard Turner</u>, Visiting Researcher; <u>Paris Perdikaris</u>, Principal Research Manager

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The possibilities of generative AI

- Data generation
- As an assistant
- How we interact with code
- Artificial General Intelligence?




Examples of Applied Research Opportunities to Implement Solutions

Kate Brodie

Senior Research Oceanographer, U.S. Army Corps of Engineers

Decadal Visioning Workshop: THE FUTURE OF COASTAL PROCESSES RESEARCH







DECADAL IMPROVEMENTS IN COASTAL OBSERVATIONS INCLUDING SATELLITES, IN-SITU, AND REAL TIME DATA DELIVERY

Kate Brodie

Senior Research Oceanographer & Field Research Facility Technical Manager US Army Engineer Research and Development Center











2014 FUTURE OF NEARSHORE PROCESSES RESEARCH



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Observation Recommendations (Elko et al. 2014):

Develop new sensors and observing techniques:



Expand long-term observing systems, conduct multiagency interdisciplinary field studies, and develop new citizen-science opportunities:



BIGGEST DECADAL ADVANCEMENTS





Large-scale, high-resolution coastal monitoring from satellites*



Operational use of close-range coastal imaging technology*



Sensor miniaturization, distributed sensing & in-water uncrewed systems*



Remote sensing renaissance for shortterm process-focused experiments



Expansion of existing long-term monitoring efforts



*expansion of satellites & uncrewed systems as data collection platforms, and agency operational use of imagery was not well predicted in the 2014 report

SATELLITE-BASED SHORELINE CHANGE MONITORING



1047 citations!

SCIENTIFIC REPORTS Interaction: Author Correction

OPEN The State of the World's Beaches



Coastal zones have h value and diverse eco heavily populated an located in the coastal of coastal landforms constline types, here stantial part of world and conhetus that ! squeeze Inevitably. assessments of the o spotial planning, sus impacts along high Despite the utility assessment of they available global scale have need derived u reported in literatury abarrelitars, second or bartier beaches (86% 70% of sandy shore time was primarily h Robust estimation shorehne position. I involved tradational Inhotographs or lider now prevale a powerf The method con by image analysis of a form, containing a co-detection. Having bo Facility of Civil Encir Delft, The Netherland Deift, The Nethonia Iwente, Enschece, Il a prioriendijkj@tad TO DE LA CLEVERTS LODIELEMONT DE LORDEN DE

Arjen Luijendijk 1,7, Gerben Hagenaars², Roshanka Ranasinghe^{1,4,7}, Fedor Baart², Gennadii Donchyts^{1,2} & Stefan Aarninkhof¹ Coastal zones co Despite the utility of historical shoreli captured since 1984 Predmannessel Minhelling and Bultware 103 (2019) 104536 present a global-sca therein Applying pi shoreline are sandy Contents lists available at Summering thus identified resi Analysis of the sate Environmental Modelling and Software eroding at rates exc sandy shorelines i ELSEVIER journal homepape: http://www CoastSat: A Google Earth Engine-enabled Python toolkit to extract shorelines from publicly available satellite imagery Kilian Vos , Kristen D. Splinter, Mitchell D. Harley, Joshua A. Simmons, Ian L. Turner Ware Research Libborney, School of Chril and Economoustand Stressorting, UNEW Sydney, 110 King Street, Manly Vela, NSW, 2023, Alastrika ABTICLE INFO ABSTRACT Reynolds: Google Parth Rogins CoastSar is an open-source software toollat written in Python that enables the user to obtain time series of dureline position at any sendy consultare worldwide from 20+ years (and growing) of publicly available satellite Stordine mapping imagery. The toolkit exploits the capabilities of Google Earth Engine to efficiently retrieve Landsat and Sentinel 2 Landsat Scutaud-2 mages cropped to any user-defined region of interest. The resulting images are pro-processed to remove cludy pixels and enhance spatial resolution, before applying a robust and generic shoreline detection algorithm. This Schoolson mendation shoreline detection technique combines a supervised image classification and a sub-pitel resolution border segmentation to map the position of the shoreline with an accuracy of -10 m. The purpose of CoastSet is to provide coastal managers, engineers and scientists a user-friendly and practical toolkit to minitoe and explore their coastlines. The software is freely-available on GitHub (https://wil Limm/kyns/Cautisni) and is secon panied by guided examples (Jupyter Notebook) plus step-by-step README documentation. Software availability and oceanic wind speeds (Young and Rillow, 2019), coupled with rising sea levels, suggest that coastal areas will be exposed to increasing haz-Software name: ConstSut ards in coming decades (Li et al., 2018). It is therefore tritical for coastal

managers and policy makers regulating coastal development to observe

and quantify these changes along coastlines vulnerable to extreme as

well as subtle changes in oceanographic forcing (e.g., flamard or all,

2017; Harley et al., 2017; Massellink et al., 3016). Data-scarcity, how

ever, remains an ongoing challenge as long-term coastal monitoring

programs, based on in sim measurements, are limited to only a few sites

around the world (e.g., darnard e) at. 3015; Planes et al., 2015; Turr

behaviour (Selinter et al. 2018) and calibrate numerical models capable

of predicting luture shoreline changes (Splinler et al., 2013), Publicly available remote sensing data from Earth Observation Satellites proides a low-cust solution to obtain long-term observations of coastline

changes over the past three decades at many sites worldwide. Crucially,

al, 2016). Indeed, shoreline data is key to understand past beach

Developer: Killian Vos Year first official release: 2018 Hardware requirements: PC System requirements: Windows, Linux, Mar Program language: Python Program size 1 MB Availability: https://witnib.com/knos/EcastSet License: GPL-3.0 Documentation: README in (lithub repository and guided example in the form of an editable Jupyter Notebook

1. Introduction

The coastal region is the most heavily urbanised land zone in the the advent of Guozle Earth Engine (Gurelick et al., 2017) has facilitated world (Small er of, 2017) and is regarded as a critical resource in view of access to the growing archive of publicity available satellite imagery. providing the opportunity for global-scale analyses stretching back deits recreational, environmental and economic importance. Yet, ocean cades (e.g., Donenus et al., 2016;11 or al., 2019; 100(and) a Pt a), 2018; coasts are affected by variations in mean sea level, extreme waves, storm surges and river flow through a range of physical processes (Ramanaghe, Meniactil es al., 2018). 2015 for a review). Recent intensification in mean wave energy files Space-horne observations have been employed in a wide range of mer al., 2010), extreme coastal wave energy (Mentaschi et al., 2017) change detection applications, including the analysis of meandering

* Corresponding author, Water Research Laboratory, School of Civil and Environmental Engineering, UNSW Sydney, Australia F-muil maleress: Louis Communities (R. Vos)

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Received 6 February 2019; Received in revised form 11 June 2019; Accepted 24 September 2019 Available unline 25 Seguenther 2019 1364-8152/0 2019 The Authors. Published by Elsevier had. This is an open access article under the CL HY-NC-ND license Fig. 1: Rapid evolution of satellite-derived shoreline methods.



I SATELLITE-BASED SHORELINE CHANGE MONITORING







Vos et al 2023

OTHER SATELLITE APPLICATIONS









Satellitederived topography for coastalchange monitoring (Almar et al., 2022; Brown et al., In-Prep)



UNCLASSIFIED



Satellite observations of Harmfull Algal Blooms (HABs) for water quality monitoring (Wolny et al., 2020)

2 OPERATIONAL COASTAL IMAGING





CorpsCam EXAMPLE APPLICATION



VIDEO IMAGING : HOURLY OBSERVATIONS, MAPS + ENGINEERING DATA



Funded by the USACE Coastal and Ocean Data Systems Program

UNCLASSIFIED

OPERATIONAL WAVE OBSERVATIONS

CDIP National Nearshore Buoy Network



Funded by the USACE Coastal and Ocean Data Systems Program

IMPACT: Provides highly accurate real-time and historical nearshore wave observations Nationally that fill a gap between NOAA deep water observations and the shoreline.

Operations: inform engineering analysis and design of federal projects, real-time port operations decisions, and stormimpact assessments.

Research: provide offshore boundary conditions that provide context for coastal change analysis and input/validation for nearshore & surf-zone models.





WAVE BUOY NETWORK: APPLICATION EXAMPLE



Automated storm bulletins summarize impacts; Identifies historic storm



CDIP Wave Observations: West Coast Atmospheric River January 4-6, 2023



 Waves at CDIP 254 Pt Santa Cruz reached 5.81 m Hs. in 20 m CDIP wave bulletins





USACE District identifies underwater damage that it otherwise wouldn't have known was there, demonstrating both the longterm and real-time value of buoy network.

- Without the continuous long-term observations, this historic significance of this storm would have been unknown
- Without the realtime, automated data processing of this network, the District would not have known to immediately initiate inspection of it's infrastructure.

Funded by the USACE Coastal and Ocean Data Systems Program

3 WAVE BUOY EVOLUTION



Fixed Networks





Small Drifting Buoys





Miniature Wave Following Buoys





Thompson et al., 2023

Feddersen et al., 2023

Buoys of Opportunity

JaiaBot

ONR/Penn State Eagle







Liquid Robotics Wave Glider SubUAS Naviator

3 APPLICATION: BUOYS OF OPPORTUNITY



Liquid Robotics Wave Glider





Amador et al., 2023

ONR Eagle





Figure 5. Direction-frequency spectra from the 8 m array (left) and the Eagle (right). The color scale indicates the spectral density [m²/Hz/deg] on a log scale. Collins and Brodie 2021







1 H







4 REMOTE SENSING OF COASTAL PROCESSES



\rightarrow Observations at process-relevant time-scales and spatial scales



Collaborative Experiment – September 2022

 Hydrodynamics: <u>high spatial-</u> <u>resolution wave resolving</u> <u>measurements</u> provide details of non-linear wave processes

Funded by the USACE Coastal and Ocean Data Systems, ERDC Basic Research Program, OSD (R&E) LUCI Program



4 REMOTE SENSING OF COASTAL PROCESSES





intermediate depths (Campana et al., 2016) Warm/cool rip plumes using Airborne IR imagery (Moulton et al., 2021)



X-band radar and drone internal waves and rip currents



9.5

(km)

ngs

Ź

7.5





5 EXPANSION OF LONG-TERM MONITORING DATA



→ Observations can be made routinely at process-relevant time-scales and spatial scales

- Hydrodynamics: high spatial-resolution wave resolving measurements
- Morphology: evolution at the time-scale of waves during storms; fully 3D, regional spatial scales



EXPANSION OF LONG-TERM MONITORING DATA



- → Observations can be made routinely at process-relevant time-scales and spatial scales
 - Hydrodynamics: high spatial-resolution wave resolving measurements
 - Morphology: evolution at the time-scale of waves during storms; <u>fully 3D, regional spatial scales</u>

UNCLASSIFIED







Funded by the USACE Coastal and Ocean Data Systems Program











USACE VISON FOR OPERATIONAL COASTAL MONITORING



Coastal Ocean Data Systems will provide Decision Support Tools to USACE Engineers at National scales supported by layered, operational <u>monitoring</u> and <u>modeling</u> of coastal evolution through integrated <u>big coastal data</u>.



National-scale Operational Coastal Monitoring Technology



COASTAL PROCESSES R&D → OPERATIONS ROAD MAP



R&D Focus Areas



Probabilistic Multi-Scale Modeling Capabilities

 \rightarrow Helps provide answers now & informs engineering actions.

→ Better communicates risk & uncertainty in our predictions.



Targeted Basic Research on Coastal Processes

 \rightarrow Builds new understanding to improve future answers to targeted questions

Vision: Seamless integration of Academic Basic R&D, Agency Applied & Operational Tech. Development, End-User Operations



Multi-Echelon Support (Funding support, end-user & community input, Gov agency synchronization)



Cohesive strategy and development goals to meet Tactical & Strategic Goals.



Collaborative R&D integrating modeling and observation capabilities from academic partners, government researchers, commercial entities, and end-users



Challenges

Discuss what the community needs: What challenges will we face in the next 10 years?

Co-developed Solutions

Discuss how we move forward: How can we address these challenges?





Tools and Approaches Discuss implementation strateg

What ways will we achieve the solutions?

Research Questions

Discuss the gaps: What basic research questions and data gaps need to be addressed in order for the tools and approaches to be successful?



BREAKOUT #3

Tools and Approaches to Support Solutions

How the Agenda is Built

Working through challenges and solutions to actionable research

Decadal Visioning Workshop: THE FUTURE OF COASTAL PROCESSES RESEARCH

Breakout Session Instructions

WHAT'S MY BREAKOUT ROOM?

Your badge has a star on it that corresponds to one of the following

breakout rooms:

JUNE 11 – JUNE 13

Hilton St. Petersburg Bayfront – 333 1st Street, SE

Decadal Visioning Workshop THE FUTURE OF COASTAL PROCESSES RESEARCH

Break

Decadal Visioning Workshop 2024 THE FUTURE OF COASTAL

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Decadal Visioning Workshop THE FUTURE OF COASTAL PROCESSES RESEARCH

Research Example: Sediment Transport

Greg Wilson, Associate Professor, Oregon State University

Sediment Transport

With Opinions From: Greg Wilson Based on Research With: Jack Aldrich, Alex Hay, Steve Henderson, Pat Dickhudt

US Nearshore Decadal Visioning Workshop – June 14, 2024

Outline

- 1. Research landscape theory and applications across scales
- 2. Some recent research (fun)
- 3. Challenges & research gaps
- 4. Path forward (tomorrow morning)

Scope: What is Nearshore Sediment Transport?

- 1. Mobilization of sand under nearshore waves and currents
- 2. Transport of sand, and subsequent changes to the bed
- 3. Feedbacks (coupling with nearshore hydrodynamics)

Scales of Sediment Transport

Blasstex – testing basic theories in the field

USCRP-Funded project: *Can we directly measure sediment transport processes and test models?*

MFDop Instrument:

- Brainchild of Alex Hay, Len Zedel & Nortek (~1995-present)
- Technical specs:
 - Vertical profile: z = 0 0.25 m, 4 mm bin-size
 - Sampling rate: 60 Hz
 - Approx. 1.4-2.1 MHz bandwidth
 - Up to 4 discrete frequencies (tone-burst sequence)

Blasstex-2 frame: MFDop, Aquascat (0.5,1MHz), Vector, bedscanning sonar

Blasstex Field Data

Model Validation

$$\langle C \rangle = \frac{\rho}{(\rho_s - \rho)g} \left[\frac{\epsilon_b}{\tan\phi} \langle u_*^2 \rangle + \frac{\epsilon_s}{w_s} \langle u_*^2 | u_\infty | \rangle \right]$$

- Time-dependent results
 - Black: Measured C(t)
 - Red: Model (using measured u_*)
- *Timing* of suspension events is good
- Errors come from sediment "burst" *magnitude*
 - Not every u_{*} event corresponds to a suspension event, and vice-versa

$$\langle C \rangle = \frac{\rho}{(\rho_s - \rho)g} \left[\frac{\epsilon_b}{\tan\phi} \langle u_*^2 \rangle + \frac{\epsilon_s}{w_s} \langle u_*^2 | u_\infty | \rangle \right]$$

• Zooming Out in Time: Better correspondence can be seen, on average

Model Validation

- Depth- and time-averaged sediment concentration and transport
- Data includes range of skewness, asymmetry, breaking
- After enough averaging, the model looks pretty accurate





Prospects for Observational Research

"The more we invest in funding sediment transport field measurements, the less we seem to understand about sediment transport." (Anon, ca. 90's)

- Why is this? Data suggests it's because of intermittency at small scales.
- *Lots of averaging* is required
 - Spatial: Must resolve full vertical profile
 - Time: Need to average over many waves. ~3.5x more than for wave stats
- Only recently have we been able to average through the noise, as necessary for testing theories



Sediment Transport as a Research Field

"You should work on sediment transport just as a hobby. Save your main research for something more mature and tractable." (Anon)

- What defines a viable/mature field of research?
 - 1. Rigorous theoretical basis (e.g., Navier Stokes Equations)
 - 2. Applications: Predictions of the theory (e.g., wave hydrodynamics)
 - **3**. Observations: Testing of the predictions
- When a field has these things, we have a machine for solving problems
- Does sediment transport meet these criteria?



August 1967

Longshore Current Velocity: A Review of Theory and Data

CYRIL J. GALVIN, JR.

Lack of *Abstract.* A proven prediction of longshore current velocity is not available, and reliable data on longshore currents are lacking over a significant range of possible flows. Theoretical studies have been based on over-simplified models, and empirical predictions

Theoretical studies have been based on over-simplified models, and empirical predictions have been hampered by lack of data. The empirically modified, momentum-flux theory now accepted as the best prediction is based on an untenable assumption and supported by inappropriate data. Regardless of their validity, however, all six of the testable equations agree fairly well with at least one of six sets of published data, and two agree with both of the better sets of data. These two equations may be used as empirical guides for velocity prediction in the absence of a proven theory. The best prospect for a generally valid velocity prediction appears to be an empirical correlation based on reliable data.

No rigorous theoretical basis



Theoretical Sediment Transport

We have excellent fundamental ideas to draw upon...

- *High-resolution modeling (CFD, DNS, LES, ...)*
- *Rheology of granular mixtures, phenomenological approaches*
- Stochastic advection-diffusion theory, intermittency
- Theories for emergent behavior and pattern formation
- Thermodynamics / Statistical Physics
 - Kinetic gases
 - Bagnold's energetics concept



Observational Sediment Transport

Given adequate field observations, *many* clear **testable research questions** come to mind:

- "Is there enhanced transport under asymmetric wave crests?"
- "Does sediment respond instantly to waves, or is there a phase lag?"
- "Why is it so intermittent? Do our theories predict that?"
- "Do our equations for longshore transport (CERC) hold up?"
- [m/s] [m/s] -250 150 200 300 50 100 (C_{tot}): Meas. = 3.2e-05 m, Model = 3.1e-05 m (Error = 4%) ×10⁻⁴ C_{tot} [m] 2 150 200 50 250 100 300

• etc. etc. etc.



Possible Breakout Discussions

- Teaching. How do we train (and recruit) students in sediment transport?
 - A series of practical formulas?
 - Fundamentals and theoretical underpinnings?
- What's the most optimal way to direct research funding?
 - Observational programs?
 - Theoretical development?
 - Graduate training?
- What would a "N. American School" of sediment transport look like?



Thank you

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JUNE 11 – JUNE 13

Hilton St. Petersburg Bayfront – 333 1st Street, SE

Decadal Visioning Workshop THE FUTURE OF COASTAL PROCESSES RESEARCH

Wrap up of Day 2

Linda Manning, President, The Council Oak LLC (Facilitator)



Decadal Visioning Workshop 2024 THE FUTURE OF COASTAL

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