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pyDeltaRCM: a flexible numerical delta model

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Summary

River deltas provide many societal benefits, and sustainability of these landforms may be impacted by human modification and global climate change. Reduced-complexity numerical delta models incorporate limited physical processes, allowing researchers to assess the spatiotemporal evolution of landscape response to individual processes and environmental forcings. This is useful to understand, for example, shifting delta morphology due to sea-level rise, changing vegetal cover, or flooding intensity. As a result, many numerical delta models have been proposed in the literature, and results from these studies are difficult to compare because of various design and implementation choices. *pyDeltaRCM* (`v2.x`) delivers a computationally efficient and easy-to-customize implementation of the DeltaRCM numerical model (Liang, Voller, et al., 2015), enabling comparison and reproducibility in studies of delta change due to various environmental forcings.

Statement of need

River deltas are societally important landforms because they provide arable land, deep inland ports, and are home to hundreds of millions of people globally (Edmonds et al., 2020). Existing at the interface between landmasses and water bodies, deltas are impacted by a multitude of processes arising in both of these domains. For example, changes in sediment input to the delta modulate the rate at which new land is built; similarly, rising water levels in the downstream basin create flooded land. In addition to natural processes, human landscape modification renders deltaic environments more sensitive to global climate change into the future (Paola et al., 2011). Demand to understand natural delta processes, and how these processes will respond to various environmental forcings, has led to a proliferation of numerical delta models in the literature (Overeem et al., 2005).

The DeltaRCM delta model (Liang, Voller, et al., 2015) has gained popularity among geomorphologists due to an attractive balance of computational cost, realism, and interpretability (Larsen et al., 2016). For example, studies have employed the DeltaRCM design to examine delta morphology and dynamism response to sea-level rise and regional subsidence (Liang, Van Dyk, et al., 2016; Liang, Kim, et al., 2016), as well as extended model design to simulate delta evolution with vegetation (Lauzon & Murray, 2018) and ice and permafrost (Lauzon et al., 2019; Piliouras et al., 2021). However, comparison among these studies is difficult, owing to disparate code bases, various implementation choices, lack of version control, and proprietary software dependencies.

Background

Here, version 2.x of *pyDeltaRCM* is introduced; *pyDeltaRCM* is a computationally efficient, free and open source, and easy-to-customize numerical delta model based on the original DeltaRCM design. The original DeltaRCM framework is inspired by well-understood physical phenomena, and models mass movement as a probabilistic weighted random-walk process coupled with a set of hierarchical rules; the model is extensively described in Liang, Voller, et al. (2015) and Liang, Geleynse, et al. (2015).

This same framework is the basis for *pyDeltaRCM* v2.x, with a few modifications selected only to resolve known numerical instabilities, improve computational efficiency, and support reproducible simulations. *PyDeltaRCM* depends only on common Python packages `numpy` (Harris et al., 2020), `matplotlib` (Hunter, 2007), `scipy` (Virtanen et al., 2020), `netCDF4`, `pyyaml`, and `numba` (Lam et al., 2015).

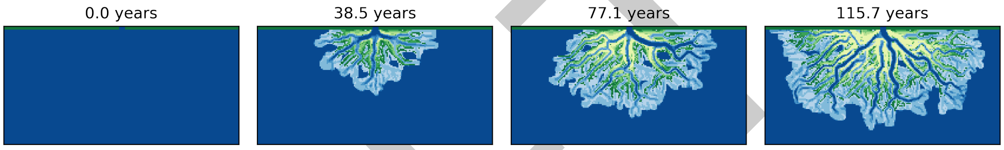

Figure 1: Simulation with *pyDeltaRCM* v2.x, default parameter set, and random `seed: 10151919`. Simulation was run for 4000 timesteps, and assumes 10 days of bankfull discharge per year.

Flexible and easy to use

pyDeltaRCM is an object-oriented package, providing the central model class `DeltaModel`. By creating custom model behavior as subclasses of `DeltaModel`, researchers can easily add, subtract, and modify model components without altering code that is not pertinent to the science objective. Importantly, separating custom code from core model code makes clear how different studies can be compared. The *pyDeltaRCM* documentation provides several examples for how to implement custom model behavior on top of the core `DeltaModel` object.

pyDeltaRCM also provides infrastructure to accelerate scientific exploration, such as the ability to configure multiple simulations from a single file. A preprocessor orchestrates `parallel` simulations for multi-core systems (optionally), and implements several tools to support simulations exploring a parameter space. For example, `matrix` expansion converts lists of parameters into an n-dimensional set of simulations. Similarly, replicate simulations can be created via an `ensemble` specification.

Reproducibility and computational efficiency were important priorities in *pyDeltaRCM* development. For example, to-disk logging records all parameters, system-level and version data, and random-seed information to ensure that all runs can be recreated. Additionally, "checkpoint" infrastructure has been added to the model, which records simulation progress during computation and can later resume model runs for further simulation. Finally, *pyDeltaRCM* uses `numba` for computational optimization (Lam et al., 2015), and does not depend on any proprietary software.

pyDeltaRCM component units and integrations are thoroughly documented and tested. Component-level documentation describes implementation notes, whereas narratives in "Guide" and "Example" documentation describes high-level model design and best practices for model use and development. *pyDeltaRCM* also couples with other numerical models via the CSDMS Basic Model Interface 2.0 (DeltaRCM Team, n.d.; Hutton et al., 2020).

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