Title: Biophysical potential and uncertainties of global seaweed farming

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The supplementary material has been appended to the end of the main manuscript.

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Biophysical potential and uncertainties of global seaweed farming

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ABSTRACT

International climate goals require over 5 gigatons/year (Gt/year) of CO₂ to be removed from the atmosphere by midcentury. Macroalgae mariculture has been proposed as a strategy for such carbon dioxide removal (CDR). However, the global potential for seaweed cultivation has not been assessed in detail. Here, we develop and use a dynamic seaweed growth model, the Global MacroAlgae Cultivation MODeling System (G-MACMODS), to estimate potential yields of four different types of seaweed worldwide, and test the sensitivity of these estimates to uncertain biophysical parameters under two nutrient scenarios (one in which the surface ocean nutrient budget is unaltered by the presence of seaweed farms, and another in which seaweed harvest is limited by nutrients that are resupplied by vertical transport). We find that 1 Gt of seaweed carbon could be harvested in 0.8% of global exclusive economic zones (EEZs; equivalent to ∼1 million km²) if farms were located in the most productive areas, but potential harvest estimates are highly uncertain due to ill-constrained seaweed mortality and nitrogen exudation rates. Our results suggest that seaweed farming could produce climate-relevant quantities of biomass carbon and highlight key uncertainties to be resolved by future research.
forcing) and intrinsic factors (biological parameters; e.g., growth rates, nutrient uptake and storage, exudation, and mortality, among others). To test sensitivities and evaluate uncertainties, we performed ~800 simulations of global growth and harvest for four seaweed types (using biophysical characteristics based on currently-farmed temperate and tropical red and brown genera). Each simulation sampled from a uniform distribution of parameter values spanning the full range of relevant values reported in the literature (Table 1), and was forced with temperature, solar irradiance, current velocities, wave height, wave period, and nutrient data sourced from a combination of satellite measurements (MODIS) and global ocean model simulations (HYCOM and CESM). Although we tested the model with forcing data from different years, results reported here reflect the year 2017 (a recent year without strong El Niño/La Niña anomalies; Supplementary Figs. 2-3), and a seasonally-variant climatology of nutrient inputs (Supplementary Fig. 4). Simulations that use parameter values best supported by literature are termed "standard runs." Seeding and harvesting for each seaweed type were optimized based on the standard runs. We also assess the importance of different model parameters via Monte Carlo methods and “random forest” classification analysis.

G-MACMODS assumes nitrogen is the limiting nutrient (i.e. implying that micronutrient constraints could be overcome by farming practices). The 800 simulations of each seaweed type were split between two bounding nutrient scenarios: (1) an "ambient nutrient" case in which average nitrate concentrations within the top 20 m are available to seaweed without depletion or competition, and (2) a "flux-limited" case where only the mass of nitrate replenished through vertical flux across 100-m depth is available to seaweed. The ambient scenario, while unrealistically optimistic for intensive production on a global scale without artificial upwelling, is illustrative of farming at a scale that does not generate substantial feedback modifying regional nutrient budgets. In contrast, the flux-limited scenario may better reflect nutrients in a situation of dense farming or nutrient competition from phytoplankton. However, both are idealized scenarios because the “offline” implementation of G-MACMODS cannot explicitly account for feedback to nutrient cycling; the different scenarios are intended to help gauge the sensitivity of seaweed production to nutrient constraints. Our analysis focuses on offshore production, as competing uses and poor resolution of coastal nutrient inputs limit model fidelity in the nearshore. The purpose of this work is not to advocate for the widespread deployment of seaweed farms over a significant fraction of the global oceans, as we expect this would come with unacceptable trade-offs to ocean health, but rather to assess the geographic distribution and potential of offshore seaweed farming to produce biomass at climate-relevant scales.

Global Seaweed Yields

Maps in Figure 1 show the magnitude and types of seaweed harvested in our standard simulations of the ambient and flux-limited nutrient scenarios (where the seaweed type that produces the largest harvest in each grid cell is farmed). Seaweed could be harvested over large areas of the ocean (208 million km$^2$ and 132 million km$^2$ in the ambient and flux-limited runs, respectively; cf. 6, 31); however, yields vary substantially in space, and annual harvests are vastly different in the two nutrient scenarios. The most productive locations include the equatorial Pacific and upwelling regions (e.g., along coasts or near energetic western boundary currents). Almost no seaweed is harvested in either nutrient scenario in the oligotrophic regimes characteristic of the center of the subtropical oceanic gyres (Figs. 1b and 1c).

Although G-MACMODS does not dynamically represent the interaction between farmed seaweed and phytoplankton, we compare the modeled rates of carbon fixation by seaweed (seaweed net primary productivity (NPP)) with phytoplankton NPP estimated from satellite ocean-color observations (Fig. 1a and Figure 1d). While phytoplankton NPP includes a significant component fueled by recycled nutrients in the euphotic zone, it represents an upper bound on new production or, similarly, net community production (NCP; typically, ~10-20% of phytoplankton NPP). Seaweed have average carbon-to-nitrogen ratios (C:N) of ~20:1 in temperate regions and ~40:1 in tropical regions, which are much higher than the ~6.6:1 (Redfield ratio) typical of phytoplankton. For the same amount of nitrogen, therefore, seaweed can fix 3-6 times as much carbon. However, in our ambient nutrient simulations, seaweed NPP is 7 and 14 times larger than observed phytoplankton NPP (~35 and 70 times larger than phytoplankton NCP) near the temperate and tropical regions, respectively, implying that the modeled seaweed growth consumes more than 10 times the nitrogen that is taken up by phytoplankton NCP (Fig. 1a). This suggests that the ambient nutrient case does not provide a sound basis for estimating potential productivity of widespread, intensive farming in the absence of artificial upwelling, but it might provide a reasonable estimate of the potential harvests of operations small enough in scale so as to not radically alter local nutrient budgets. Indeed, the yields simulated in the ambient nutrient scenario results agree well with harvest values reported in the literature for many small farms and a few large farms situated near nutrient outflows (Supplementary Figs. 5-8). In contrast, zonally-averaged seaweed NPP is less than observed phytoplankton NPP in our flux-limited simulations, except in equatorial regions where phytoplankton growth is iron-limited (Fig. 1d), consistent with our NCP constraint. The lower harvests estimated in the flux-limited scenario may therefore better reflect production when farming at scales large enough to significantly deplete the surface fixed-nitrogen inventory, relying on the influx of "new"
Figure 1. Global seaweed harvest. Maps of annual potential harvest per unit area (yield; b-c) of the preferred seaweed (the type with the largest harvest in each grid cell; f-g). White boxes correspond to regions depicted in Figure 2. Zonally-averaged annual harvest for the preferred seaweed group, seaweed net primary productivity (NPP), and phytoplankton NPP estimated from satellite observations are shown in (a,d). Zonally-averaged annual harvests for the four seaweed types are shown in (e,h).

The standard simulations of both nutrient scenarios predict that temperate brown and tropical red seaweed out-compete temperate red and tropical brown seaweeds over most of the global ocean. When nutrients are abundant, temperate red seaweed appear at the equatorward edges of regions with temperate brown seaweed (Figs. 1f and 1g). The zonally-integrated annual harvest of tropical red seaweed is 3-4 times higher than that for tropical brown seaweed; similarly, the zonally-integrated annual harvest of temperate brown seaweed is 4-8 times larger than that for the temperate reds (Figs. 1e and 1h).

At regional scales (e.g., areas enclosed by boxes in Figures 1b and 1c), physical processes such as western-boundary current meanders (Fig. 2a), coastal upwelling (Fig. 2l), and frequent eddy activity (Fig. 2k) influence environmental variability and seaweed growth. Four factors govern seaweed growth rate in the model: water temperature, nutrient availability, light, and seaweed density, or “crowding” (equation 7). Of these factors, water temperature largely determines the latitudinal distribution of different seaweed types (e.g. tropical seaweeds in the South/East China Sea (Fig. 2, top row) and temperate seaweed in the Norwegian Sea (Fig. 2, third row)). At smaller scales, nutrient availability controls regional patterns of seaweed productivity and, as expected, is more important in flux-limited simulations than in the ambient nutrient scenario (Fig. 2). Light availability and crowding (e.g. self-shading, sub-grid scale nutrient competition) can become relatively important growth limitation factors in regions with readily available nutrients.

Uncertainty Analysis

We assess the sensitivity of our results to uncertainty in the biophysical parameters in G-MACMODS, conducting a Monte Carlo analysis over a range of literature-based parameter values with uniform distribution (Table 1). The standard deviation of Monte Carlo simulations increases in direct proportion to the simulated harvest yield (Fig. 3). For example, regions with larger harvests in our standard simulations also show greater variability in the Monte Carlo results (Figs. 3e and 3f; Figures 3a and 3b as compared to maps in Figures 1b and 1c). In the most productive 10% of the regions of the ocean, the average yield can range 646–1589 GtC/km².

Based on a random forest analysis of Monte Carlo results, the biological parameters that most influence harvested seaweed yield globally are the mortality rate and nitrogen exudation rate (Figs. 3g and 3h). Our Monte Carlo simulations evaluate mortality rates from 0.003/day - 0.017/day; some prior models have used similar or slightly lower values (0.001/day - 0.01/day)28,29,43,44. Nitrogen exudation rates are more important in determining harvest in the flux-limited simulations than in the ambient scenario, since exudation (i.e., slimy excretion from the seaweed) leads to loss of nutrients that are already scarce in the
Figure 2. Regional harvest. (Maps a-d, i-l) Annual harvest yields for the boxed regions in Figure 1. (Bars e-h, m-p) Relative influence of growth parameters (equation 7) in determining regional harvest yield for each seaweed type. (Spark lines) Relative spatially integrated annual harvest for each seaweed type.

Scaling production in EEZs

The maps in Figure 4 show the area of exclusive economic zones (EEZs) that would be required to grow seaweed biomass of 1, 2, and 4 GtC/year in our standard, flux-limited simulation. Cumulative distributions of seaweed-based annual harvest in the standard, flux-limited simulation as a function of EEZ area (sorted by harvest yield, such that the areas with the largest harvests are cultivated first; Figure 4e) show diminishing returns from farming more than ~15% of EEZs (locations scattered across the world), with harvests approaching a limit of ~4.5 GtC/year at ~25% of EEZs (Fig. 4e). In the standard, flux-limited simulation, 1 GtC/year could be harvested from the most productive ~0.8% of EEZs (1 million km²; located in the equatorial Pacific; Figure 4a), ranging from 0.36 to 1.8 GtC/year at the 5th to 95th percentiles of flux-limited Monte Carlo simulations but always less than the 2.4 GtC/year yield predicted in the standard, ambient nutrient simulation (Fig. 4f).

Implications for seaweed CDR

This work represents an advance over previous estimates in that it employs a mechanistic seaweed growth model (G-MACMODS) to dynamically simulate four types of seaweed under two bounding nutrient scenarios and evaluates parametric sensitivities; it is an important first step towards a fully prognostic model. The standard simulation model results have been...
**Figure 3. Harvest uncertainty.** Maps of standard deviation from the Monte Carlo results (a,b) and probability density function (PDF) of the standard run annual harvest yield (c,d). The y-axis has been cut off to better visualize the smaller PDF values (corresponding to larger harvests). Bin-averages of Monte Carlo statistics are shown as a function of the standard run results (e,f). The median harvest is shown as a solid line; the dark and light shading denote the values between the 25th and 75th percentiles and the 5th and 95th percentiles, respectively. The dashed 1:1 line shows where the median harvest would lie if it equaled the standard harvest. The relative importance of the biological parameters in Table 1, as quantified by random forest analysis, are depicted in (g,h). $V_{max}^*$ [$\mu$mol-N/(m$^2$ h)] is the product of the maximum uptake rate ($V_{max}$) and the ratio of biomass-to-surface area (B:SA). The biological parameters not explicitly named are grouped under the "other" category (Supplementary Fig. 9).

The ambient nutrient scenario, which assumes that nutrient levels are unaffected by seaweed farms, represents a global extrapolation of current-scale, coastal seaweed farming. But it is not clear that depleted nutrients could be replaced through transport from the surrounding environment without quickly straining the inventory of global nutrients and disrupting the natural biological carbon pump\(^{31}\). Sustaining levels of production in ambient simulations over large areas would thus require some form of nutrient amendments (e.g. artificial upwelling), which would, in turn, entail additional costs. Our flux-limited nutrient simulations instead reflect offshore seaweed production that might be sustainable given local resources by using only "new" nitrogen replenished from the deep ocean (fluxed upward across the 100 m depth). Relative to the ambient scenario, the standard flux-limited simulations reduce potential seaweed harvest worldwide by an average 90%.
Figure 4. Total potential harvest in EEZs. (a-d) Areas of exclusive economic zones (EEZs) required to harvest 1, 2, and 4 GtC/year of seaweed biomass in standard, flux-limited simulations, sorted by productivity (i.e. prioritizing the most productive areas). (e-f) Cumulative distribution functions of total seaweed carbon harvested relative to the share of global EEZs farmed. Results from the ambient and flux-limited standard runs are depicted as dashed lines. The solid green line and surrounding shading indicate the range of harvests of Monte Carlo, flux-limited simulations.

Even in flux-limited simulations, though, we estimate that a climate-relevant mass of carbon (e.g., 1 GtC) might be harvested by farming seaweed in the most productive 0.8% of EEZs (Fig. 4f). However, 0.8% of EEZs worldwide (~1 million km²) would represent a roughly 370-fold increase in the area where seaweed is currently farmed (~2700 km²) [18, 45]. For comparison, the area occupied by all agricultural cropland in the U.S. is ~1.6 million km² [46]. The National Academy of Sciences suggests that if seaweed cultivation comprises one of several CDR strategies, this industry would only need to extract ~0.03 GtC/year [5]; however, even this target requires increasing the current seaweed cultivation area by over 10-fold and moving it to the most productive regions of the ocean. While conversations center on seaweed harvest yield, the harvested biomass accounts for an average 45% of the total seaweed biomass produced in our standard runs; the remaining 55% is grazed, remineralized, buried in situ, or exported from the farm as particulate organic carbon. Of the seaweed carbon that is not harvested, if 1% and 2% is buried in the shelf or exported to the deep sea, respectively (as estimated for wild seaweed [25]), then for every 1 GtC harvested from farms over the continental shelf, 0.02 GtC could be buried, and 0.03 GtC could be exported to the deep sea. However, carbon removed from the atmosphere may be less than harvested carbon due to time scales of air-sea carbon fluxes and disruptions to the natural biological carbon pump [31, 47].

As indicated by the variance among Monte Carlo simulations, the largest uncertainties in our estimates of seaweed harvest correspond to mortality and exudation rates. The mortality rate in the model reflects erosion, dislodgement, pests, herbivory, diseases, sedimentation, and natural disasters. Our standard simulations assume a mortality rate of 1% per day (Table 1).
Existing models and observations span both lower\textsuperscript{28,29,31,44,48} and higher mortality rates\textsuperscript{36,49,50}, yet these sources, which primarily consider nearshore farms, may have limited applicability to mortality on open ocean farms. Like mortality, nitrogen exudation by seaweeds is understudied, despite its importance in modeling productivity in nutrient-limited waters. Nitrogen exudation rates between 0.002/day\textsuperscript{36} and 0.2/day\textsuperscript{51} have been reported. We assume a constant rate of nitrogen exudation (0.05/day in the standard simulations), but the rate is likely to fluctuate in time with environmental conditions\textsuperscript{52–54} and ratios of nutrients in the seaweed (as observed for carbon exudation)\textsuperscript{54–57}. Although not represented in our model, exudation rates may be related to seaweed growth rates\textsuperscript{51}, and mortality rates\textsuperscript{58}. Maximum growth rate, maximum uptake rate, and half saturation constant also affect estimated harvests to varying but lesser degrees (Fig. 3). Maximum growth rate cannot be easily parsed from observations of relative growth rate, and existing maximum uptake rate and half saturation constants may not have been estimated using standardize environmental conditions. Our results thus highlight the importance of further research to narrow uncertainties related to mortality and exudation rates under real-world conditions expected during cultivation and thereby narrow the uncertainty bounds around our harvest estimates.

Despite the limitations of our model and substantial uncertainties related to biophysical parameters, we estimate the global potential for seaweed farming in unprecedented detail. Our results suggest that it may be possible to annually harvest seaweed containing 1 GtC/year by farming on the order of 1 million km\textsuperscript{2} of the most productive ocean areas. However, in addition to narrowing uncertainties and accounting for the effects of climate change, future work must further assess the economic and political feasibility of farming seaweed over such large areas that may have other uses or protections (e.g., fishing, shipping traffic, marine protected areas). Similarly, if the purpose of harvesting such large quantities of seaweed is to sink it to the deep ocean and thereby sequester carbon, the effects on abyssal ecosystems\textsuperscript{59–61} and possibility of increasing the extent of hypoxic regions\textsuperscript{52,63} deserve more investigation. But although there remain many unknowns and hurdles for large-scale seaweed farming, our analysis suggests that harvesting quantities of seaweed that would make a substantial contribution to global CDR is possible and future investment in research is warranted.

Online Methods

G-MACMODS Overview

The Global Macroalgae Cultivation Modelling System (G-MACMODS) used in this study draws on recent work on within-farm biophysics\textsuperscript{32}, using elements from previously published research\textsuperscript{26–28}. The state variables in the model are seaweed biomass (B; g-DW/m\textsuperscript{2}; where DW is dry-weight) and nitrogen cell quota (Q; mg-N/g-DW,\textsuperscript{64}). Nitrogen is the limiting macronutrient in G-MACMODS. Though we recognize that other macronutrients and micronutrients could further limit our results in, for example, high-nitrogen low chlorophyll environments\textsuperscript{42}, we assume that the aquaculture industry has implemented micronutrient fertilization. G-MACMODS estimates seaweed biomass in units of dry weight; biomass is converted to units of carbon by assuming that carbon constitutes 30% of the seaweed dry weight for all seaweed groups\textsuperscript{65,66}, though carbon content may actually be lower for tropical red seaweed\textsuperscript{39}.

A diagram of the conceptual model is presented in Supplementary Figure 1. The model has a daily time step and considers macroalgae to be grown at 2 m depth below the surface for the purposes of light attenuation. Seaweed biomass is depth-integrated across the top 20 m of the water column.

Model structure

Temporal changes in the state variables (B and Q) can be described with the following equations:

\[
\frac{dQ}{dt} = V - \mu(Q - Q_{min}) - E(Q - Q_{min}),
\]

(1)

and

\[
\frac{dB}{dt} = \mu B - d_{M}B,
\]

(2)

where V is the nitrogen uptake rate [\(\mu\)mol-N/(g-DW h)], E is a fractional exudation rate (1/day), \(\mu\) is the fractional growth rate (1/day), and \(d_{M}\) is the fractional death rate (1/day).
Nitrogen Uptake

The rate of nitrogen uptake by seaweed is determined by extrinsic (environmental) and intrinsic (biological) limiting factors:

\[ V = V_{\text{max}} f(Q) f(\|v\|, T_w, C), \]  

(3)

where \( V_{\text{max}} \) is the maximum uptake rate (Table 1), \( f(Q) \) represents a dynamic nutrient cell quota which allows for luxury uptake of nutrients, and \( f(\|v\|, T_w, C) \) represents both kinetic and mass-transfer limitations on nitrogen uptake. We use a linear nutrient cell quota:\(^2\):

\[ f(Q) = \frac{Q_{\text{max}} - Q}{Q_{\text{max}} - Q_{\text{min}}}, \]  

(4)

where \( Q_{\text{min}} \) is the minimum amount of nitrogen that should be found in a seaweed cell (structural nitrogen), \( Q_{\text{max}} \) is the maximum amount of nitrogen stored internally, such that uptake decreases as the internal nitrogen concentration increases, and \( f(Q) \) is a unitless coefficient between 0 and 1. The parameter \( f(\|v\|, T_w, C) \) in equation (3) is a limit on uptake based on a combination of Michaelis-Menten kinetics and mass-transfer limitation regulated by the surrounding waves and currents:\(^67\text{-}\text{69}\):

\[ f(\|v\|, T_w, C) = \frac{C}{K_m \left( \frac{C}{K_m} + \frac{1}{2} \left( \gamma + \sqrt{\gamma^2 + 4 \frac{C}{K_m}} \right) \right)}, \]  

(5)

where \( \gamma = 1 + (V_{\text{max}} / \beta K_m) - (C / K_m), K_m \) is the half-saturation constant (Table 1), \( C \) is the external concentration of nitrogen, and

\[ \beta = \frac{D}{\delta_D} + \frac{4 \delta_D}{T_w} \sum_{n=1}^{\infty} \left( 1 - \exp \left( \frac{-Dn^2 \pi^2 T_w}{2 \delta_D^2} \right) \right), \]  

(6)

with units of m/s. In equation (6), \( D \) is the molecular diffusivity of nitrate at 18\(^\circ\)C (7.3 \times 10^{-10} \text{ m}^2/\text{s})\(^32\text{,70}\), \( T_w \) is wave period, and \( \delta_D \) is the thickness of the diffusive boundary layer, defined using the thickness of the viscous boundary layer \( \delta_D = \delta_v = 10 \nu / (\sqrt{\nu} |v|) \) where \( \nu \) is the molecular kinematic viscosity (10\(^{-6}\) \text{ m}^2/\text{s}) and \( C_D \) is the drag coefficient\(^69\) (Table 1). The parameter \( f(\|v\|, T_w, C) \) is unitless and varies between 0 and 1. Note that this nitrogen uptake model assumes that (a) the diffusion boundary layer is completely stripped away every half a wave period, regardless of the size of the wave, (b) the thickness of the viscous boundary layer \( (\delta_v) \) can be parameterized with the thickness of the viscous boundary layer \( (\delta_v) \), and (c) that we can ignore near-boundary turbulent transport (i.e. assume the blade is smooth)\(^69\), though this has been shown to enhance exchange rates\(^71\). We do not consider within-canopy flow reduction, which negatively affects uptake\(^32\text{,72}\). We assume that wave height has a negligible affect on uptake, since renewal of the diffusive boundary layer (and, hence, enhanced nutrient uptake) can occur through blade flapping in low-flow environment\(^73\). Thus, equation (3) is used to estimate the amount of nitrogen that the seaweed could, theoretically, absorb from the environment (\(dN\)).

Two nutrient scenarios are tested in this study: (1) a case where nutrient concentrations are averaged over the top 20 m of each grid cell and are available to seaweed without depletion or competition is referred to as the "ambient nutrient" scenario, and (2) a case where the amount of nutrients available for uptake is capped by the nitrogen fluxed upward through the 100-m depth plane \( (N_{\text{new}}) \), referred to as the "flux-limited" scenario. In the flux-limited scenario, the nitrogen uptake rate (equation 3) is still determined by the "ambient" (average of top 20 m) nutrient concentration, but if the amount of nitrogen that would be theoretically taken up by seaweed at a given time is greater than that fluxed upward at 100 m depth, \( dN > N_{\text{new}} \), then uptake (\( V \) in equation 1) is capped using \( dN = N_{\text{new}} \). Additional simulations were performed to test an alternate depth for estimating \( N_{\text{new}} \) - at the annual maximum mixed-layer depth at each grid cell - but resulting productivity differences were relatively small compared to other uncertainties presented in the Uncertainty Analysis section (median increase of 5% in the annual harvest yield).

Growth

Similar to the nitrogen uptake rate, growth rate (\( \mu \)) is also constrained by extrinsic and intrinsic limiting factors:
\[
\mu = \mu g(k) g(Q) g(T) g(E),
\]

(7)

where \(\mu g(k)\) (1/day) is the maximum growth rate at a given seaweed density, accounting for the crowding effects of self-shading and within-farm (sub-gridscale) nutrient limitation. The maximum growth rate is further constrained by the internal nitrogen cell quota \(g(Q)\), water temperature \(g(T)\), and light \(g(E)\), all of which are unitless coefficients, varying between 0 and 1.

The growth rate limitation imposed by crowding in the seaweed canopy embodies the general idea that less-dense seaweed can grow faster, described as

\[
\mu g(k) = A B^{-0.75},
\]

(8)

where \(A \left[ (1/d)/(g-DW/m) \right]\) is a factor that represents the growth rate at the maximum allowable biomass density. Strictly defined, \(A = k_{R}/B_{cap}^{-0.75}\), where \(B_{cap}\) (g-DW/m) is the maximum biomass density and \(k_{R}\) is the maximum growth rate at \(B_{cap}\) [chosen to be 5% per day and tuned to match values documented in the literature for our different seaweed types (Supplementary Figs. 5 - 8)]. The power law in equation (8) was derived by re-fitting data from a comprehensive meta analysis74. Our new fit was applied over the data in ref. 15 and binned to 0.01-width bins from 0-1 g/L and 0.1-width bins for 1-60 g/L seaweed density, weighted by the number of observations in each bin (with a minimum weight of 8 observations). Our fit excluded data corresponding to total-nitrogen (NO\(_3\)+NH\(_4\)) conditions not likely to be found in the surface ocean (values above 20 \(\mu\)M). Although according to equation (8), \(\mu g(k) \rightarrow \infty\) as \(B \rightarrow 0\), we cap the maximum growth rate \(\mu_{max}\) according to values found in the literature (Table 1), such that \(\mu g(k) \rightarrow \mu_{max}\) as \(B \rightarrow B_{seed}\), where \(B_{seed}\) is the seed weight.

The nitrogen quota limitation \(g(Q)\) in equation (7) follows the Droop model64:

\[
g(Q) = \frac{Q - Q_{min}}{Q}.
\]

(9)

where \(Q_{min}\) is set per species type (Table 1). The temperature limitation term in Equation (7) is similar to a Gaussian probability curve75:

\[
g(T) = \exp\left( -\beta_1 (T - T_{opt})^2 \right), \quad T < T_{opt}
\]

\[
g(T) = \exp\left( -\beta_2 (T - T_{opt})^2 \right), \quad T > T_{opt}
\]

\[
g(T) = 1, \quad T = T_{opt},
\]

(10)

where \(T_{opt}\) is a 5\(^\circ\)C optimal temperature range for each seaweed group that we are examining. \(T\) is the daily temperature, and the \(\beta_1\) and \(\beta_2\) coefficients are adjusted to reach zero near the lower and upper temperature limits, respectively.

The light-limitation in equation (7) is largely informed by phytoplankton studies76:

\[
g(E) = f \frac{I - I_c}{I_e - I_c} \exp\left( \frac{I - I_c}{I_e - I_c} + 1 \right),
\]

(11)

where \(I_e\) and \(I_c\) are the daily-averaged saturating and compensating irradiance (W/m\(^2\)), \(f\) is the fraction of daylight that is implemented to account for periods of darkness, and \(I\) is the irradiance reaching an underwater depth of 2 m. The irradiance is attenuated following the implementation in the Marine Biogeochemistry Library (MRBL)77,78.

**Mortality**

The mortality rate, \(d_M\) in equation (2), is the sum of a constant daily mortality rate that is meant to incorporate grazing, aging, and disease \(d\) (Table 1) and a term that accounts for breakage from waves \(d_w\), such that \(d_M = d + d_w\). The \(d_w\) term is dependent on wave power and, as such, is variable in both time and space79:

\[
d_w = (2.3 \times 10^{-4})(P \times 10^3) + 2.2 \times 10^{-3},
\]

(12)
where \( P \) is wave power in Watts:

\[
P = \frac{\rho g^2 H_s^2 T_w}{64\pi}
\]  

(13)

where \( \rho \) is the density, \( H_s \) is the significant wave height, and \( T_w \) is the wave period.

**Environmental data**

The environmental inputs applied to our model (water temperature, solar irradiance, current velocities, wave height, wave period, and nutrient concentrations) stem from a combination of satellite measurements and global ocean model outputs spanning multiple years. For the purposes of this manuscript, we explore a suite of simulations using inputs from 2017, the most recent year with available data that is also not identified with having a strong ENSO index. Input data from 2003-2019 were used in simulations examining inter-annual differences in estimated seaweed productivity (Supplementary Fig. 10), however, regional inter-annual variability was comparatively small with respect to parameter uncertainty and is therefore not the focus of this study.

Sea surface temperature (SST) and surface photosynthetically active radiation (PAR) are used as a proxy for in – situ temperature and irradiance, respectively, over the depth of macroalgae growth. SST and PAR used in this study are 8-day averages from the MODerate Resolution Imaging Spectroradiometer (MODIS; R2018), on the NASA Earth Observing System, with spatial resolution of 1/12°. Net oceanic primary production (NPP) was estimated from MODIS chlorophyll measurements using the Vertically Generalized Production Model (VGPM)\(^{34}\). SST, PAR, and NPP were downloaded from the Ocean Productivity website (https://sites.science.oregonstate.edu/ocean.productivity/index.php).

Zonal and meridional current velocities were extracted from the HYbrid-Coordinate Ocean Model (HYCOM\(^{89}\)) Global Ocean Forecasting System (GOFS) 3.1, accessed from https://www.hycom.org/dataserver/gofs-3pt1/analysis. HYCOM is a global data-assimilating model\(^{81}\) with 1/12° horizontal resolution and 41 depth levels, of which we use the surface velocities.

Significant wave height and wave period were taken from the European Centre for Medium-Range Weather Forecasts (ECMWF) ERA5\(^{82}\) atmospheric reanalysis produced by the Copernicus Climate Change Service\(^{83}\). ERA5 provides hourly significant wave height of combined wind waves and swell, and mean wave period with a 1/2° horizontal resolution. The data are averaged to 8-day time intervals.

Nutrient information is taken from a high resolution biogeochemical simulation led by the National Center for Atmospheric Research (NCAR) and run in the Community Earth System Model (CESM) framework\(^{34,85}\). The biogeochemical model has a 1/10th° horizontal resolution and 62 depth levels. Fields used in this study include 5-day mean nitrate concentrations averaged over the upper 20 meters, and vertical fluxes of nitrate across the 100-m depth plane were calculated to provide and estimate of fluxes of new nitrogen into the euphotic zone. All 5-day outputs were interpolated to 8-day periods for consistency with the other environmental inputs to G-MACMODS.

Although G-MACMODS steps forward with a daily time step, we apply the 8-day environmental inputs that best correspond to the G-MACMODS time stamp. All environmental inputs were spatially interpolated onto a 1/12° global grid, using linear interpolation if the input data were of higher resolution, or nearest-neighbor if the input data were of lower resolution.

**Seaweed groups**

Here, we focus on four seaweed groups containing seaweed species that are among the world’s ten most cultivated by weight\(^{86}\): tropical reds (e.g. *Eucheuma, Gracilaria, Kappaphycus*), tropical browns (e.g. *Sargassum*), temperate reds (e.g. *Porphyra*), and temperate browns (e.g. *Saccharina, Laminaria, Macrocystis*). Values of parameters required by G-MACMODS were gathered from available literature for a few representative seaweed genera (Table 1); “standard runs” were defined using average (when multiple parameter estimates were available) or speculated values (based on information from other genera when there were few or no published values). We define the temperature parameters in equation 6 similarly, using available information for representative genera (Table 2). The optimal temperature range in equation 6 is extended to a 5° width, rather than a single number, to account for variations within a seaweed genus.

The standard runs were spun up for one year, and the seeding was optimized by choosing the run initialization date that yielded the largest yearly biomass harvest (averaged across 2003-2019) for every grid point. Tropical and temperate brown
seaweed runs were seeded with 50 g-DW/m². Tropical and temperate red seaweed runs were seeded with 200 g-DW/m² and 10 g-DW/m², respectively, following examples in the literature (see references in Supplementary Figs. 5-8). Seaweed are seeded with an initial nitrogen cell quota ($Q_0$), such that

$$Q_0 = Q_{\text{min}} + \frac{N}{35} (Q_{\text{max}} - Q_{\text{min}}),$$

(14)

where $N/35$ is the ratio of the ambient nitrogen concentration at the time of seeding to the a representative N concentration below the nutricline (35 µM).

Model-Field Data Comparison

To test our choice of standard parameters (Table 1) and calibrate $B_{\text{cap}}$ and $k_R$ (equation 8), harvested yield from our standard runs was compared to literature values of harvested yield that encompass ocean-cultivated and wild seaweed stocks. Only farmed values published after the year 2000 are included to account for changes in technology and methods across the years, whereas we include wild stock values from literature published as far back as 1990.

To test our model performance around tropical red seaweed, we executed ambient runs with a 45-day harvest period (standard in literature; e.g. 87–92) and compared the maximum amount of biomass harvested at every grid cell within a single harvest period to *Eucheuma* and *Kappaphycus* harvest yields in the literature (Supplementary Fig. 5). The median harvest yield in G-MACMODS is larger than the median harvest yield reported in the literature, but the 50% range surrounding the medians is comparable. Our model never reaches some of the larger harvest values in the literature, but we attribute that to the difference in farming depths; eucheumoids are typically farmed in depths shallower than 10 m, often very close to shore where terrestrial nutrient inputs may be significant, whereas our model considers nutrients depth-averages over the top 20 m of the water column.

G-MACMODS performance for tropical browns, temperate reds, and temperate browns was tested without including harvest, comparing the maximum seaweed biomass per grid cell in our ambient runs to the maximum biomass per unit area harvested on a seaweed plot or observed standing stock (Supplementary Figs. 6-8). All median harvest yields from G-MACMODS surpass the median harvest yields in the literature, suggesting that the G-MACMODS results are optimistic.

Harvest

Harvest schemes were based on available information of current farming practices (e.g. 19, 31, 91–95) and optimized for each seaweed group to achieve maximal biomass per harvest based on standard run tests of three harvest schemes: periodic harvesting, periodic harvesting with a biomass threshold, and conditional harvesting (with a dual criteria of a target weight or when death exceeds growth). The test runs also allowed for optimization of the target weight to initiate harvest (10%, 20%, 30%, 40%, 50%, or 80% of $B_{\text{cap}}$), as well as the percent of biomass removed at each harvest (40%, 60%, or 80%). Finally, the number of harvests per year were limited based on documented cultivation practices. The temperate brown and red alga are commonly harvested twice¹⁹ and 6 times a year³⁹, respectively, while the tropical brown and red alga are harvested up to 8 times a year⁹²,⁹⁶. Temperate brown seaweeds were allowed to grow without consideration for harvest for at least 60 days after seeding. Considering the above factors, the harvesting schemes that produced the highest harvested yields for each seaweed group are as follows:

1. Tropical red and brown seaweeds: Harvest occurs every 45 days only if the seaweed biomass has reached the target weight of 800 g-DW/m² (27% of $B_{\text{cap}}$) for tropical reds and 400 g-DW/m² (50% of $B_{\text{cap}}$) for tropical browns. If 45 days elapse and the seaweed does not reach its target weight, another 45-day period must transpire before re-evaluating the biomass. If the biomass has reached its target weight, then 80% of the biomass is harvested.

2. Temperate red seaweeds: Harvest is initiated whenever the biomass reaches the target weight (80 g-DW/m², 40% of the $B_{\text{cap}}$) within 150 days after seeding or if the death exceeds growth for 7 days. If the biomass has reached its target weight then 80% of the biomass is harvested; if the death exceeded growth for > 7 days or the final harvest period is reached, 99% of the biomass is harvested (1% loss rate assumed in final total harvest).

3. Temperate brown seaweeds: Harvest occurs when the biomass reaches the target weight (1350 g-DW/m², 68% of the $B_{\text{cap}}$) within 220 days after seeding or if death exceeds growth for 7 days. If the biomass has reached its target weight, then 80% of the biomass is harvested; if the death exceeded growth for > 7 days or the end of 220 days is reached, 99% of the biomass is harvested (1% loss rate assumed in final total harvest).
Monte Carlo simulations

We used Monte Carlo methods to estimate the uncertainty surrounding our standard run harvest amounts. We performed between 425 - 450 Monte Carlo simulations for each seaweed group and nutrient scenario (ambient vs. flux-limited). Each Monte Carlo simulation chose the value of the seaweed biological parameters using a uniform probability distribution bounded by the magnitudes in Table 1. When possible, these bounds are 25% greater (lower) than the maximum (minimum) biological parameter values found in the literature. The mean, median, standard deviation, and percentiles (5th, 25th, 75th, 95th) of annual harvest yields resulting from these Monte Carlo simulations were calculated across each model grid cell. The relative importance of each Monte Carlo parameter value upon harvested biomass was evaluated using random forest analysis.

Model Limitations

G-MACMODS and our scenarios are subject to a number of important limitations and caveats. First, neither of the implemented nutrient scenarios consider how seaweed farms affect the surrounding hydrodynamics, which can substantially affect nutrient uptake and yields but are challenging to resolve in a global-scale model. Moreover, the nutrient data (from CESM simulations) do not resolve nutrient runoff in coastal areas, sources of nitrogen other than nitrate (e.g., ammonia or urea), nor consider other limiting macronutrients such as phosphate. These nutrient-related limitations may affect our harvest estimates in specific locations, and perhaps lead us to underestimate harvest in some nearshore areas. On the other hand, operating farms will not have the benefit of hindsight that our model uses to optimize seeding and harvest schedules, and the model assumptions are optimistic with regard to micronutrient fertilization and environment/strain optimization in cultivars. G-MACMODS would also benefit from a more refined expression of seaweed mortality that could account for episodic events (e.g. storms, diseases) and nonlinear grazing pressure, among other factors. Finally, we do not explicitly model the effects of climate change and projected changes in ocean conditions that can stress growing seaweeds, shift their geographical distribution, increase the frequency and severity of storms, decrease nutrient fluxes by enhanced stratification, and make diseases and epiphytes more prevalent. The are important areas for future research. Although certainly not a proxy for the many effects of climate change, we note that interannual variability in environmental forcing 2003-2019 affects our harvest estimates less than the uncertainties related to biological parameters (Supplementary Fig. 10).

Table 1. Biological Parameters

| Parameter                  | Genus   | Standard Values | Monte Carlo Bounds | References |
|----------------------------|---------|-----------------|--------------------|------------|
| $V_{\text{max}}$ (µmol-N/(g-DW h)) | *Eucheuma* | 9.7             | [4.05, 16.3]        | 100, 101   |
| Maximum uptake rate        | *Sargassum* | 17.9            | [1.86, 36.9]       | 102, 103   |
|                            | *Porphyra* | 52.2            | [26.3, 90]         | 104, 105, 106, 107 |
|                            | *Macrocystis* | 12.8           | [2.3, 38.1]      | 108, 109, 110, 111, 112 |
|                            | *Saccharina* | 11.8           | [1.9, 30]          | 113, 29    |
| $K_{\text{m}}$ (µM)        | *Eucheuma* | 5.6             | [0.2, 13.8]        | 100        |
| Half-saturation constant   | *Sargassum* | 3.0             | [1.1, 5.5]         | 103        |
|                            | *Porphyra* | 5.2             | [1.5, 12.7]        |            |
|                            | *Macrocystis* | 10.1          | [3.2, 18.1]      | 108, 109, 111 |
|                            | *Saccharina* | 2              | [1.1, 4.2]         | 114, 113   |
| $\mu_{\text{max}}$ (1/day) | *Eucheuma* | 0.2             | [0.1, 0.3]         | 115, 116, 117 |
| Maximum growth rate        | *Sargassum* | 0.2             | [0.1, 0.3]         | 118, 119   |
|                            | *Porphyra* | 0.2             | [0.1, 0.3]         | 104, 120   |
|                            | *Macrocystis* | 0.2           | [0.1, 0.3]        | 121, 122   |
|                            | *Saccharina* | 0.2           | [0.1, 0.3]        | 123, 124, 125, 126 |
| $\text{PAR}_{s}$ (µmol photon/(m² s)) | *Eucheuma* | 125.9           | [52.1, 550]        | 127, 128, 129, 130, 131 |
| Saturating irradiance     | *Sargassum* | 303.9          | [112.5, 643.8]     | 119, 132   |
| Parameter                      | Genus         | Standard Values | Monte Carlo Bounds | References  |
|-------------------------------|---------------|-----------------|--------------------|-------------|
|                               | *Porphyra*    | 104             | [34.5, 233.8]      | 133, 134    |
|                               | *Macrocystis* | 212.4           | [105.8, 350]       | 135, 136    |
|                               | *Saccharina*  | 76.3            | [11.3, 212.5]      | 137, 138, 113 |
| PAR<sub>c</sub> (µmol photon/(m<sup>2</sup> s)) | *Eucheuma*   | 13.5            | [3.8, 32.5]        | 127, 128, 129, 130, 131 |
| Compensating irradiance       | *Sargassum*  | 26              | [3.8, 46.3]        | 119, 132    |
|                               | *Porphyra*    | 24.8            | [6.8, 54]          | 133, 134    |
|                               | *Macrocystis* | 20.5            | [7.5, 43.1]        | 135, 136    |
|                               | *Saccharina*  | 15.5            | [5.7, 29.3]        | 137, 113    |
| Q<sub>min</sub> (mg-N/g-DW)   | *Eucheuma*   | 5.8             | [4.3, 7.2]         | 139         |
| Minimum nitrogen cell quota   | *Sargassum*  | 5.8             | [4.3, 7.2]         | 139         |
|                               | *Porphyra*    | 10.2            | [7.6, 12.7]        | 139         |
|                               | *Macrocystis* | 10.2            | [7.6, 12.7]        | 139         |
|                               | *Saccharina*  | 10.2            | [7.6, 12.7]        | 139         |
| Q<sub>max</sub> (mg-N/g-DW)   | *Eucheuma*   | 44              | [33, 55]           | 139         |
| Maximum nitrogen cell quota   | *Sargassum*  | 44              | [33, 55]           | 139         |
|                               | *Porphyra*    | 54              | [40.5, 67.5]       | 139         |
|                               | *Macrocystis* | 54              | [40.5, 67.5]       | 139         |
|                               | *Saccharina*  | 54              | [40.5, 67.5]       | 139         |
| C<sub>D</sub> (unitless)      | *Eucheuma*   | 0.5             | [0.01, 1]          | 139         |
| Drag coefficient              | *Sargassum*  | 0.5             | [0.01, 1]          | 139         |
|                               | *Porphyra*    | 0.5             | [0.01, 1]          | 139         |
|                               | *Macrocystis* | 0.5             | [0.01, 1]          | 139         |
|                               | *Saccharina*  | 0.5             | [0.01, 1]          | 139         |
| B:SA (g-DW/m<sup>2</sup>)     | *Eucheuma*   | 94.8            | [71.1, 118.5]      | L. Roberson (personal comm.) |
| Ratio of biomass to surface area | *Sargassum* | 333             | [249.8, 416.3]     | 29          |
|                               | *Porphyra*    | 10              | [7.5, 12.5]        | 108, 111, 110, 141 |
|                               | *Macrocystis* | 58              | [43.5, 72.5]       | 108, 111, 110, 141 |
|                               | *Saccharina*  | 58              | [43.5, 72.5]       | 108, 111, 110, 141 |
| E (1/day)                     | *Eucheuma*   | 0.05            | [0.001 0.1]        | 36, 43, 51  |
| Exudation                     | *Sargassum*  | 0.05            | [0.001 0.1]        | 36, 43, 51  |
|                               | *Porphyra*    | 0.05            | [0.001 0.1]        | 36, 43, 51  |
|                               | *Macrocystis* | 0.05            | [0.001 0.1]        | 36, 43, 51  |
|                               | *Saccharina*  | 0.05            | [0.001 0.1]        | 36, 43, 51  |
| d (1/day)                     | *Eucheuma*   | 0.01            | [0.003 0.03]       | 28, 29, 36, 43, 79 |
| Death rate                    | *Sargassum*  | 0.01            | [0.003 0.03]       | 28, 29, 36, 43, 79 |
|                               | *Porphyra*    | 0.01            | [0.003 0.03]       | 28, 29, 36, 43, 79 |
|                               | *Macrocystis* | 0.01            | [0.003 0.03]       | 28, 29, 36, 43, 79 |
|                               | *Saccharina*  | 0.01            | [0.003 0.03]       | 28, 29, 36, 43, 79 |
Table 2. Temperature Parameters used in Equation 6

| Genus       | $T_{opt}$ (°C) | $\beta_1$ | $\beta_2$ | References       |
|-------------|----------------|-----------|-----------|------------------|
| *Eucheuma*  | 22.5-27.5      | 0.09      | 0.09      | 142–144          |
| *Sargassum* | 22.5-27.5      | 0.09      | 0.09      | 145, 146         |
| *Porphyra*  | 12-17          | 0.03      | 0.09      | 134, 147         |
| *Macrocystis* | 13-18      | 0.04      | 0.05      | 148, 149         |
| *Saccharina* | 10-15         | 0.03      | 0.1       | 124, 126, 138    |
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Author contributions statement

K.A.D., C.F., B.S., and S.D. conceived the work. I.B.A.S. wrote the first draft off the manuscript. I.B.A.S., C.F., and B.S. designed and coded G-MACMODS, with the help of K.A.D. B.S. downloaded and interpolated the environmental forcing data. M.L. provided the nutrient data. M.L., S.D., and J.D. provided context for the atmospheric and economic implications of the work. All authors contributed to interpreting the results, as well as framing and revising the paper.
Biophysical potential and uncertainties of global seaweed farming - Supplementary Material

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Supplementary Figures

Figure 1. Schematic overview of G-MACMODS. For details, please refer to the Online Methods section of the manuscript.
Figure 2. Satellite-derived variables. Temporal mean and standard deviation of the sea surface temperature (SST; top row), surface irradiance (PAR; middle row), and phytoplankton net primary productivity (NPP; bottom row) stemming from MODIS.

Figure 3. Hydrodynamic variables. Temporal mean and standard deviation of the significant wave height and mean wave period from ECMWF (top and middle row, respectively), as well as the surface current speed from HYCOM (bottom row).
Figure 4. CESM NO$_3$. Temporal mean and standard deviation of the CESM depth-average ambient NO$_3$ concentrations (top row) and NO$_3$ flux across the 100-m depth plane (bottom row).
Figure 5. Model-field comparisons (tropical red seaweed). Locations of farmed *Eucheuma* and *Kappaphycus* observations in 1–19 (left panel). Some neighboring locations may not be resolved (may be plotted on top of other locations). Boxplots of *Eucheuma* and *Kappaphycus* harvest in 1–19, as well as the maximum tropical red seaweed biomass harvested in a single harvest cycle in G-MACMODS ambient nutrient simulations (right panel). Only values above B = 200 g-DW/m² (the tropical red seed weight in G-MACMODS) are shown in the boxplots (n = 74 values from 19 articles and > 600,000 values from G-MACMODS). Pink triangles indicate the mean harvest value in the literature articles referenced above. Values reported in fresh weight were converted to dry weight by assuming a dry-to-wet biomass ratio of 1:10.
Figure 6. Model-field comparisons (tropical brown seaweed). Locations of wild Sargassum observations in 20–30 (left panel). Boxplots of wild Sargassum standing stock in 20–30, as well as the maximum tropical brown biomass observed in G-MACMODS ambient nutrient simulations when harvest is not imposed (to better match the wild seaweed values in the literature) are shown in the right panel. Only values above $B = 50 \text{ g-DW/m}^2$ (the tropical brown seed weight in G-MACMODS) are shown in the boxplots ($n = 40$ values from 10 articles and $> 900,000$ values from G-MACMODS). Green circles indicate the mean reported biomass in the literature articles referenced above. Values reported in fresh weight were converted to dry weight by assuming a dry-to-wet biomass ratio of 1:10.
Figure 7. Model-field comparisons (temperate red seaweed). Locations of farmed Pyropia observations in 7, 12, 31–33 (left panel). Boxplots of Pyropia harvest in 7, 12, 31–33, as well as the maximum temperate red seaweed biomass harvested in a single harvest cycle in G-MACMODS ambient nutrient simulations (right panel). Only values above B = 10 g-DW/m² (the temperate red seed weight in G-MACMODS) are shown in the boxplots (n = 53 values from 5 articles and > 1.8 million values from G-MACMODS). Pink triangles indicate the mean harvest values in the literature articles referenced above. Values reported in fresh weight were converted to dry weight by assuming a dry-to-wet biomass ratio of 1:10.
Figure 8. Model-field comparisons (temperate brown seaweed). Locations of wild (34–43; green circles) and farmed (44–56; pink triangles) *Saccharina, Laminaria*, and *Macrocystis* observations (left panel). Some neighboring locations may not be resolved (may be plotted on top of other locations). Boxplots of wild standing stock values from 34–43 (n = 127 values from 10 articles), harvest from 44–56 (n = 80 values from 13 articles), and maximum kelp biomass output from G-MACMODS (ambient nutrient scenario) when harvest is not imposed (n > 1.5 million values) are shown in the right panel. Only values above B = 50 g-DW/m$^2$ (the temperate brown seed weight in G-MACMODS) are shown in the boxplots. Green circles and pink triangles indicate the mean reported biomass in the literature articles that discuss wild and farmed kelp, respectively. Values reported in fresh weight were converted to dry weight by assuming a dry-to-wet biomass ratio of 1:10.
Figure 9. Random forest results. Focused view of the "other" category in Figure 3 of the main manuscript. The results have been normalized to reflect the relative importance of each parameter. For parameter information, please refer to Table 1 of the main manuscript.
Figure 10. Inter-annual harvest variability. Harvest yield temporal mean (top row) and standard deviation (bottom row) across 2002–2019.
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Supplementary Figures

Figure 1. Schematic overview of G-MACMODS. For details, please refer to the Online Methods section of the manuscript.
Figure 2. Satellite-derived variables. Temporal mean and standard deviation of the sea surface temperature (SST; top row), surface irradiance (PAR; middle row), and phytoplankton net primary productivity (NPP; bottom row) stemming from MODIS.

Figure 3. Hydrodynamic variables. Temporal mean and standard deviation of the significant wave height and mean wave period from ECMWF (top and middle row, respectively), as well as the surface current speed from HYCOM (bottom row).
Figure 4. CESM NO$_3$. Temporal mean and standard deviation of the CESM depth-average ambient NO$_3$ concentrations (top row) and NO$_3$ flux across the 100-m depth plane (bottom row).
Figure 5. **Model-field comparisons (tropical red seaweed).** Locations of farmed *Eucheuma* and *Kappaphycus* observations in 1–19 (left panel). Some neighboring locations may not be resolved (may be plotted on top of other locations). Boxplots of *Eucheuma* and *Kappaphycus* harvest in 1–19, as well as the maximum tropical red seaweed biomass harvested in a single harvest cycle in G-MACMODS ambient nutrient simulations (right panel). Only values above $B = 200 \text{ g-DW/m}^2$ (the tropical red seed weight in G-MACMODS) are shown in the boxplots ($n = 74$ values from 19 articles and > 600,000 values from G-MACMODS). Pink triangles indicate the mean harvest value in the literature articles referenced above. Values reported in fresh weight were converted to dry weight by assuming a dry-to-wet biomass ratio of 1:10.
Figure 6. Model-field comparisons (tropical brown seaweed). Locations of wild Sargassum observations in 20–30 (left panel). Boxplots of wild Sargassum standing stock in 20–30, as well as the maximum tropical brown biomass observed in G-MACMODS ambient nutrient simulations when harvest is not imposed (to better match the wild seaweed values in the literature) are shown in the right panel. Only values above $B = 50 \text{ g-DW/m}^2$ (the tropical brown seed weight in G-MACMODS) are shown in the boxplots ($n = 40$ values from 10 articles and $> 900,000$ values from G-MACMODS). Green circles indicate the mean reported biomass in the literature articles referenced above. Values reported in fresh weight were converted to dry weight by assuming a dry-to-wet biomass ratio of 1:10.
Figure 7. Model-field comparisons (temperate red seaweed). Locations of farmed *Pyropia* observations in 7, 12, 31–33 (left panel). Boxplots of *Pyropia* harvest in 7, 12, 31–33, as well as the maximum temperate red seaweed biomass harvested in a single harvest cycle in G-MACMODS ambient nutrient simulations (right panel). Only values above B = 10 g-DW/m$^2$ (the temperate red seed weight in G-MACMODS) are shown in the boxplots (n = 53 values from 5 articles and > 1.8 million values from G-MACMODS). Pink triangles indicate the mean harvest values in the literature articles referenced above. Values reported in fresh weight were converted to dry weight by assuming a dry-to-wet biomass ratio of 1:10.
Figure 8. Model-field comparisons (temperate brown seaweed). Locations of wild (34–43; green circles) and farmed (44–56; pink triangles) *Saccharina*, *Laminaria*, and *Macrocystis* observations (left panel). Some neighboring locations may not be resolved (may be plotted on top of other locations). Boxplots of wild standing stock values from 34–43 (n = 127 values from 10 articles), harvest from 44–56 (n = 80 values from 13 articles), and maximum kelp biomass output from G-MACMDS (ambient nutrient scenario) when harvest is not imposed (n > 1.5 million values) are showin in the right panel. Only values above B = 50 g-DW/m² (the temperate brown seed weight in G-MACMDS) are shown in the boxplots. Green circles and pink triangles indicate the mean reported biomass in the literature articles that discuss wild and farmed kelp, respectively. Values reported in fresh weight were converted to dry weight by assuming a dry-to-wet biomass ratio of 1:10.
Figure 9. Random forest results. Focused view of the "other" category in Figure 3 of the main manuscript. The results have been normalized to reflect the relative importance of each parameter. For parameter information, please refer to Table 1 of the main manuscript.
Figure 10. **Inter-annual harvest variability.** Harvest yield temporal mean (top row) and standard deviation (bottom row) across 2002–2019.
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