# Modelling Plankton Dynamics in the Oceanic Mixed Layer

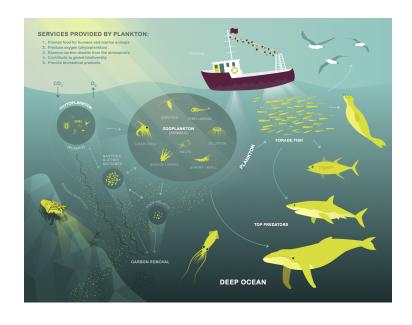
Lecture 5

Ago Merico

29 January 2020

Part 1 – Main concepts and NPZ model

Part 2 - Physical dynamics and NPZD model



#### **Nutrients**

#### The fundaments of life...

Inorganic

Macro-

Carbon (C)
Nitrogen (N)

Phosphorus (P) Silicon (Si)

. . .

Micro-

Iron (Fe) Cobalt (Co)

Manganese (Mn)
Copper (Cu)

Zinc (Zn)

\_.... (\_.

Organic

Carbohydrates Lipids

Proteins

Vitamins

. . .

Photosynthesis: uses inorganic nutrients and light to synthesise biomass

$$\underbrace{106\,\text{CO}_2}_{\text{carbon dioxide}} + \underbrace{16\,\text{HNO}_3 \, + \, \text{H}_3\text{PO}_4}_{\text{nutrients}} + \underbrace{122\,\text{H}_2\text{O}}_{\text{water}} \quad \rightleftarrows \quad \underbrace{1(\text{CH}_2\text{O})_{106}\,(\text{NH}_3)_{16}\,(\text{H}_3\text{PO}_4)}_{\text{organic matter}} + \underbrace{138\,\text{O}_2}_{\text{oxygen}}$$

#### Bacteria

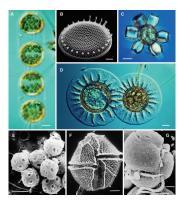
- Most diverse groups of organisms in the marine realm;
  - many are free-living, others form symbiotic associations;
  - many thrive within dead organic substrates;
  - important recycling role.
- Can live under extreme conditions;
  - e.g. high salinity, high temperature, anoxic waters.
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#### **Phytoplankton**

- Phytoplankton is a diverse group of microscopic, mostly single-celled, photosynthetic organisms that drift with the currents in marine and fresh waters.
- Although accounting for less than 1 % of Earth's photosynthetic biomass, phytoplankton are responsible for more than 45 % of our planet's annual net primary production.
- They are agents for primary production, the creation of organic compounds from CO<sub>2</sub> dissolved in the water, a process that sustains the aquatic food web and life in general.
- Phytoplankton depend on inorganic nutrients, primarily nitrate, phosphate, and silicic acid, whose availability is governed by a balance between the biological pump and upwelling of deep, nutrient-rich waters. Across large regions of the World Ocean phytoplankton are also limited by lack of iron.



Examples of representative marine phytoplankton. (A) Diatom chain; (B) single valve diatom; (C) coccolithophore; (D) overlapping pair of phycomas; (E) clump of coccospheres; (F) dinoflagellate; (G) dinoflagellate. Scale bars: (A, C, E, F) 10  $\mu$ m; (B and G) 2  $\mu$ m; and (D) 25  $\mu$ m.

# Zooplankton

- Grazers: dominant consumers of phytoplankton;
- Transporters: eat at one depth, extrete at another;
- Through their consumption and processing of phytoplankton and other food sources, zooplankton play a key role in aquatic food webs, as a resource for consumers on higher trophic levels (e.g. fish).



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- to investigate the effects of possible future events, like the doubling of atmospheric CO<sub>2</sub>, that cannot be experimentally determined;
- a standard problem in marine ecology is the causal explanation of the phytoplankton spring bloom in temperate regions;
- there are benthic models, fishery models, population-dynamical models for whales, and many more...

Phytoplankton spring blooms are a common feature of temperate coastal areas; the sequence of events are as follows:

 phytoplankton stock is low through winter because strong vertical mixing keeps net losses from the photic zone greater than net growth, despite sufficient nutrient; low sun angles and short days contribute to keep stock low by limiting algal growth rates:

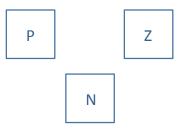
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- as spring gives way to summer, the growth of phytoplankton depletes nutrients that
  are no longer rapidly supplied from depth because of water stratification; productivity becomes nutrient limited and declines; also grazing reduces algal growth rates,
  and sinking of algal cells due to nutrient starvation produce a mid summer low in
  algal stocks;

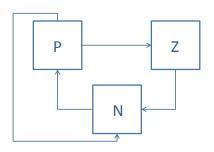
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- 5. the fall bloom is mixed away by the storms of early winter, which also resupply the surface with nutrients; the cycle begins again.

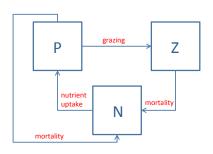
The basic goals in marine ecology are to understand the interactions among nutrient avialability, phytoplankton growth, phytoplankton stock size and zooplankton stock size.



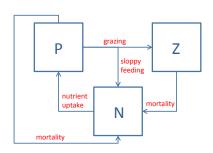
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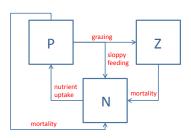


The differential equations can be written in words as follows:

$$\frac{dP(t)}{dt} = + Nutrient \ Uptake - Mortality_P - Grazing ,$$

$$\frac{dZ(t)}{dt} = + (Assimilation \ Efficiency) * Grazing - Mortality_Z \,,$$

$$\frac{dN(t)}{dt} = -Nutrient \ Uptake + (1 - Assimilation \ Efficiency) * Grazing + Mortality_P + Mortality_Z$$
.



#### **Processes**

The processes involved in the interactions among the different dynamic variables (also called compartments) can be spelled out as follows:

Nutrient Uptake 
$$= \mu(N) P = \mu_{max} \left( \frac{N}{N + K_N} \right) P$$
, 
$$\text{Grazing} = g(P) Z = g_{max} (1 - e^{\Lambda P}) Z ,$$
 
$$\text{Mortality}_P = m_P P ,$$
 
$$\text{Mortality}_Z = m_Z Z ,$$
 Assimilation Efficiency  $= \gamma$ .

with: 
$$m_P = 0.1 \,\mathrm{d}^{-1}$$
;  $m_Z = 0.2 \,\mathrm{d}^{-1}$ ;  $\gamma = 0.3$  (i.e. 30%)

#### Nutrient uptake

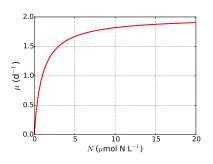
#### Nutrient uptake is simulated with a Monod-type function:

$$\mu(N) = \mu_{max} \left( \frac{N}{N + K_N} \right)$$

with:

$$\mu_{max} = 2.0 \text{ d}^{-1} \text{ (maximum growth rate)}$$

$$K_{N}=1.0~\mu \mathrm{mol}~\mathrm{N}~\mathrm{L}^{-1}$$
 (half-saturation constant)



## Grazing

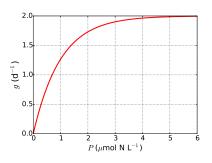
#### Grazing is simulated using the Ivlev's formulation:

$$g(P) = g_{max}(1 - e^{-\Lambda P})$$

with:

$$g_{max} = 2.0.5 \text{ d}^{-1}$$
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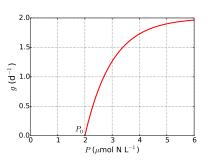
$$g(P) = g_{max}(1 - e^{-\Lambda (P - P_0)})$$

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$$\Lambda = 1.0 \ \mu \text{mol N L}^{-1} \ \text{(Ivlev constant)}$$

$$P_0=2.0~\mu\mathrm{mol}\,\mathrm{N}\,\mathrm{L}^{-1}$$
 (minimum grazing threshold)



## Complete model equations

This model was developed by Franks et al. (1986), the complete equations are:

$$\begin{split} \frac{dP(t)}{dt} &= \mu_{\text{max}} \left( \frac{N}{N + K_N} \right) P - \textit{m}_P \, P - \textit{g}_{\text{max}} \left[ 1 - e^{-\Lambda (P - P_0)} \right] \, Z \,, \\ \frac{dZ(t)}{dt} &= \gamma \, \textit{g}_{\text{max}} \left[ 1 - e^{-\Lambda (P - P_0)} \right] \, Z - \textit{m}_Z \, Z \,, \\ \frac{dN(t)}{dt} &= -\mu_{\text{max}} \left( \frac{N}{N + K_N} \right) P + \textit{m}_P \, P + \textit{m}_Z \, Z + (1 - \gamma) \, \textit{g}_{\text{max}} \left[ 1 - e^{-\Lambda (P - P_0)} \right] \, Z \,. \end{split}$$

## Complete model parameters

Symbol	Description	Value	Unit
$\mu_{ extit{max}}$	maximum growth rate (P)	2.0	$d^{-1}$
$K_N$	half-saturation constant	1.0	$\mu \mathrm{mol}\mathrm{N}\mathrm{L}^{-1}$
g <sub>max</sub>	maximum ingestion rate $(Z)$	1.5	$d^{-1}$
٨	lvlev constant	1.0	$\mu \mathrm{mol}\mathrm{N}\mathrm{L}^{-1}$
$P_0$	minimum grazing threshold	0.0	$\mu \mathrm{mol}\mathrm{N}\mathrm{L}^{-1}$
$m_P$	mortality rate $(P)$	0.1	$d^{-1}$
$m_Z$	mortality rate $(Z)$	0.2	$d^{-1}$
$\gamma$	assimilation efficiency	0.3	fraction

with: 
$$P(0) = 0.3 \, \mu \mathrm{mol} \, \mathrm{N} \, \mathrm{L}^{-1}, \ Z(0) = 0.1 \, \mu \mathrm{mol} \, \mathrm{N} \, \mathrm{L}^{-1}, \ N(0) = 1.6 \, \mu \mathrm{mol} \, \mathrm{N} \, \mathrm{L}^{-1}$$

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For consistency, also the parameters of the model are expressed in units of nitrogen, wherever relevant;

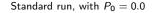
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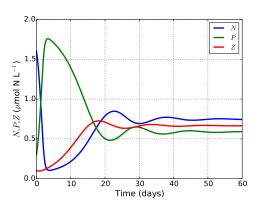
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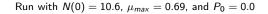
In this way nitrogen is said to represent the *currency* of the model.

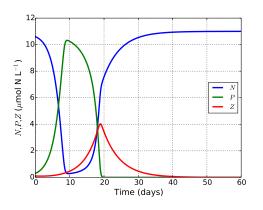
#### Results of the NPZ model



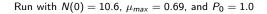


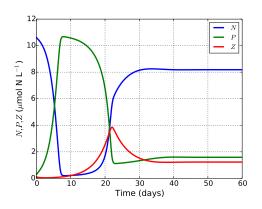
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# A (somewhat) more realistic NPZ model

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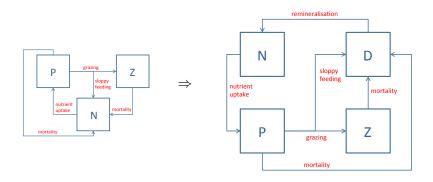
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- Thus a more realistic model would simulate control of primary production by seasonally varying sunlight and include the decrease of light with depth, at least to the bottom of a mixing layer;
- Phytoplankton growth varies non-linearly with irradiance (P-I relationship), and irradiance decreases exponentially downward, so one should integrate production down through the water column; variations in mixing should be also included (at least in a simplified fashion) for example by varying the mixed-layer depth through the simulated season.

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- Zooplankton can be taken to sustain their stock within the mixed layer by swimming;
- ▶ We add another dynamic variable, 'detritus' (D), to reflect the fact that organic matter (fecal materials and dead organisms) is temporarily stored in this compartment where it gets decomposed by bacteria and then transferred to restock the inorganic nutrient pool.



Adding the detritus pool implies that (1) all dead material pass by this compartment before reaching the inorganic nutrient pool and (2) a new process must be added, *remineralisation*, which represents the decomposition (or break down) of organic materials into inorganic minerals by bacteria.

# End of part 1

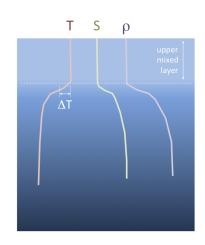
Part 2

## Ocean upper mixed layer

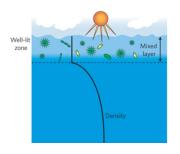
Turbulence create a well mixed surface layer where temperature (T) salinity (S) and density  $(\rho)$  are nearly uniform with depth;

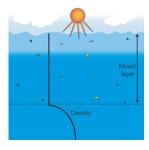
The depth of this mixed layer is measured by the depth at which T is some value less than SST (e.g.  $\Delta T = 0.5$ );

Undergoes large seasonal changes;



## Upper mixed layer depth dynamics

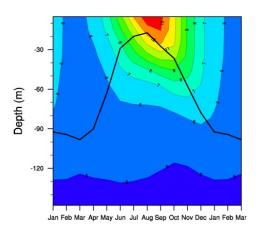




When the mixed layer is shallow, phytoplankton mostly stay within the well-lit zone, where they thrive. When the mixed layer deepens, phytoplankton are distributed over a larger volume of water, and sink more easily to depths too dark to sustain them.

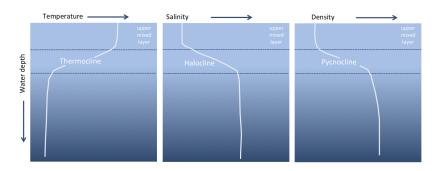
# Seasonal cycle of MLD and T

Northeast Pacific (50 °N, 145 °W)



## Below upper mixed layer

Just below the upper mixed layer, there is another water layer where temperature, salinity, and density change rapidly; these layers are called, respectively, thermocline, halocline and pycnocline.



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- Such slab models can be run quickly and straightforwardly, enabling both a multitude of runs and ease of analysing results;
- Despite the simplicity of the two-layer slab physics, these models are sufficiently well formulated to permit realistic and insightful simulations of marine ecosystems.

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In the case of non-motile entities, a deepening of the mixed layer (i.e. when the rate of change of M is positive, dM/dt>0) will dilute the concentration (thus decreasing P). In contrast, when the mixed layer shallows, material is left behind, or detrained, but the concentration in the mixed layer will remain unchanged because the loss due to detrainment is balanced by the increased phytoplankton density due to the decreased volume (thus detrainment, i.e. negative rate changes of M, dM/dt < 0, will not alter the concentration of P).

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Diffusive mixing across the thermocline is parameterised with a constant factor  $(\kappa)$ .

For non-motile entities, the whole diffusion term is, thus, written as

$$K = \frac{\kappa + h^+}{M(t)}$$

For motile entities (i.e. zooplankton), the diffusion term is simply

$$\frac{h(t)}{M(t)}$$

Light limitation:

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Phytoplankton growth is limited by light availability. This requires the consideration of (1) photosynthetically active radiation (PAR) on the surface of the ocean (i.e. I at z=0,  $I_0$ ), (2) attenuation of PAR with depth z, and (3) photosynthesis as a function of light. PAR is attenuated with depth according to the Beer-Lambert law with attenuation coefficient  $k_{PAR}$ :

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thus the light limiting term is given by:

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Instead of solving the integral numerically at each model time step, we will use the following analytical solution (according to Anderson et al. 2015):

$$\Psi(I) = \frac{1}{M(t)} \int_{0}^{M} P(I) dz = \frac{P_{max}}{k_{PAR} M} \ln \left( \frac{\alpha I_{0} + \sqrt{P_{max}^{2} + (\alpha I_{0})^{2}}}{\alpha I_{M} + \sqrt{P_{max}^{2} + (\alpha I_{M})^{2}}} \right)$$

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with  $\alpha$  initial slope of the P-I curve,  $P_{max}$  maximum photosynthetic rate (i.e. the phytosynthetic rate at  $I\to\infty$ ),  $I_0$  irradiance on the top surface (i.e. at z=0), and  $I_M$  irradiance at z=M (i.e.  $I_M=I_0\,\mathrm{e}^{-k_{PAR}\,M}$ ) with M mixed layer depth.

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 $k_{PAR}$  is the sum of attenuation due to water and phytoplankton, parameters  $k_w$  and  $k_p$ , respectively:

$$k_{PAR} = k_w + k_p P$$

Temperature dependence:

#### Temperature dependence:

Phytoplankton growth depends on temperature according to the Eppley's formulation:

$$f(T) = e^{0.063 T}$$

with  $T=\mbox{Sea}$  Surface Temperature or SST.

Phytoplankton gross growth:

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Phytoplankton gross growth is the product of light-dependent growth with different limiting terms (varing between 0 and 1):

Gross growth = 
$$\underbrace{e^{0.063 T}}_{\text{temp dependence}} \cdot \underbrace{\left(\frac{N}{N + K_N}\right)}_{\text{nutrient limitation}} \cdot \underbrace{\psi(I)}_{\text{light limitation}}$$

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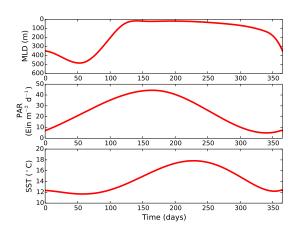
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Temporal changes in temperature (T), irradiance at the top of the ocean surface  $(I_0)$ , and mixed layer depth (M) drive the seasonal changes in the plankton ecosystem model.

These environmental data are external input to the model and represent the *physical forcing* to the model. Thus data files for T,  $I_0$ , and M are required; these files are called SST (seas surface temperature), PAR (irradiance at z=0), and MLD (mixed layer depth), respectively.

# Physical forcing

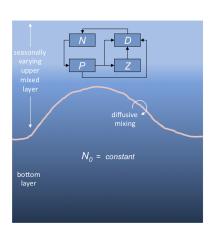
Environmental data typical of a temperate region of the North Atlantic:



# Physical scheme – summary

#### Zero-dimensional "slab" model:

- The water column is devided into two layers, which are assumed to be homogeneously mixed;
- 2. Lateral advection is not considered:
- Environmetal data are external inputs and used as physical forcing to the model;
- Biological and ecological processes occur only in the upper mixed layer;
- Nutrient concentration is assumed constant in the bottom layer.



Grazing by zooplankton:

### Grazing by zooplankton:

Grazing by zooplankton is assumed to be on both phytoplantkon and detritus. This choice illustrates how to implement ingestion on multiple prey types. Our grazing functions reflect a passive switching response (according to Anderson et al. 2015) and are formulated as follows:

$$G_P = \left(\frac{g_{max}\,\varphi_P\,P^2}{K_Z^2 + \varphi_P\,P^2 + \varphi_D\,D^2}\right)$$

$$G_D = \left(\frac{g_{\text{max}} \, \varphi_D \, D^2}{K_Z^2 + \varphi_P \, P^2 + \varphi_D \, D^2}\right)$$

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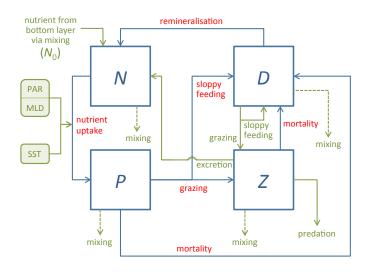
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These are sigmoidal functions and the terms  $g_{max}$ ,  $K_Z$ ,  $\varphi_P$ , and  $\varphi_D$  are parameters indicating, respectively, maximum grazing (or ingestion) rate, half-saturation constant for intake, preference for phyotplankton, and preference for detritus.

# NPZD model - flow diagram



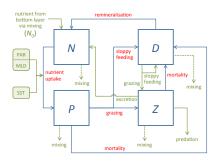
## NPZD model - equations

$$\begin{split} \frac{dN(t)}{dt} &= -f(T) \, U(N) \, \Psi(I) \, P + m_D \, D + \underbrace{\beta \, \big( 1 - K_{NZ} \big) \, \big( G_P + G_D \big) \, Z}_{\text{excretion}} + K \, \big( N_0 - N \big) \, ; \\ \frac{dP(t)}{dt} &= f(T) \, U(N) \, \Psi(I) \, P - m_P \, P - G_P \, Z - K \, P \, ; \\ \frac{dZ(t)}{dt} &= \beta \, K_{NZ} \, \big( G_P + G_D \big) \, Z - m_Z \, Z - m_Z^* \, Z^2 - \frac{h(t)}{M} \, Z \, ; \\ \frac{dD(t)}{dt} &= m_P \, P + m_Z \, Z + \underbrace{\big( 1 - \beta \big) \, \big( G_P + G_D \big) \, Z}_{\text{sloppy feeding}} - G_D \, Z - m_D \, D - K \, D \, . \end{split}$$

with:

$$\begin{split} f(T) &= e^{\mathbf{0.063}\,T}, \quad U(N) = \frac{N}{N + K_N}, \quad \text{and} \quad T = \text{sst.dat} \\ \Psi(I) &= \frac{P_{max}}{(k_W + k_P P)\,M} \, \ln \left( \frac{\alpha\,I_0 + \sqrt{P_{max}^2 + (\alpha\,I_0)^2}}{\alpha\,I_M + \sqrt{P_{max}^2 + (\alpha\,I_M)^2}} \right), \quad I_M = I_0\,e^{-(k_W + k_P P)M}, \quad \text{and} \quad I_0 = \text{par.dat} \\ K &= \frac{\kappa + h^+}{M(t)}, \quad h^+ = max(h,0), \quad h = \frac{dM}{dt}, \quad \text{and} \quad M = \text{mld.dat} \\ G_P &= \left( \frac{g_{max}\,\varphi_P\,P^2}{K_Z^2 + \varphi_P\,P^2 + \varphi_D\,D^2} \right) \quad \text{and} \quad G_D &= \left( \frac{g_{max}\,\varphi_D\,D^2}{K_Z^2 + \varphi_P\,P^2 + \varphi_D\,D^2} \right). \end{split}$$

# NPZD model – flow diagram and equations



$$\frac{dN(t)}{dt} = -f(T) U(N) \Psi(I) P + m_D D + \underbrace{\beta (1 - K_{NZ}) (G_P + G_D) Z}_{\text{excretion}} + K (N_0 - N);$$

$$\frac{dP(t)}{dt} = f(T) U(N) \Psi(I) P - m_P P - G_P Z - K P;$$

$$\frac{dZ(t)}{dt} = \beta K_{NZ} (G_P + G_D) Z - m_Z Z - m_Z^* Z^2 - \frac{h(t)}{M} Z;$$

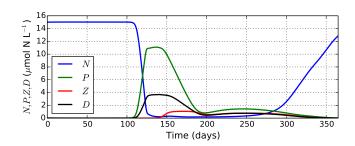
$$\frac{dD(t)}{dt} = m_P P + m_Z Z + \underbrace{(1 - \beta) (G_P + G_D) Z}_{\text{elements}} - G_D Z - m_D D - K D.$$

## NPZD model – parameters and initial conditions

Symbol	Description	Value	Unit
No	nitrogen in bottom layer	15.0	$\mu$ mol/L
$P_{max}$	maximum photosynthetic rate	1.1	$d^{-1}$
gmax	maximum ingestion rate	0.9	$d^{-1}$
$\kappa_N$	half saturation constant $(P)$	0.85	$\mu$ mol/L
$K_Z$	half saturation constant $(Z)$	0.6	$\mu$ mol/L
$m_P$	mortality rate (P)	0.2	$d^{-1}$
$m_Z$	mortality rate $(Z)$	0.1	$d^{-1}$
$m_D$	remineralisation rate $(D)$	0.6	$d^{-1}$
β	zooplankton ingestion efficiency	0.69	fraction
$K_{NZ}$	zooplankton assimilation efficiency	0.75	fraction
$k_W$	light extinction due to water	0.05	$m^{-1}$
$k_p$	light extinction due to $P$	0.03	$m^{-1}  (\mu mol/L)^{-1}$
$\alpha$	initial slope of P-I curve	0.15	$({\sf Ein}{\sf m^{-2}})^{-1}{\sf d^{-1}}$
$\kappa$	cross thermocline mixing	0.1	$\mathrm{m}\mathrm{d}^{-1}$
$m_Z^*$	mortality due to higher predators $(Z)$	0.34	$(\mu mol/L)^{-1}d^{-1}$
$\varphi_P$	grazing preference for P	0.67	%
$\varphi_D$	grazing preference for D	0.33	%

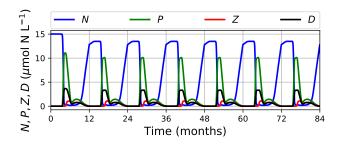
with:  $N(0) = 15.0 \, \mu \text{mol L}^{-1}$ ,  $P(0) = 0.01 \, \mu \text{mol L}^{-1}$ ,  $Z(0) = 0.01 \, \mu \text{mol L}^{-1}$ ,  $D(0) = 0.01 \, \mu \text{mol L}^{-1}$ .

### NPZD model - results



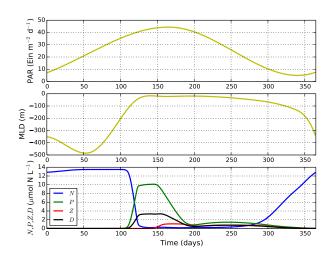
### NPZD model - results

Run the model over several years by forcing it every year with the same environmental variables to reach a stable solution after an initial spin-up phase.



### NPZD model - results

Stable solutions (4th year after spin-up phase of 3 years) plotted along with forcing variables (PAR and MLD).



## Suggestions and hints

#### Upload the external forcing files as follows:

```
mld = np.loadtxt("/Users/ago/CompLifeSci/mld.dat") # mixed layer depth (M)
par = np.loadtxt("/Users/ago/CompLifeSci/par.dat") # irradiance at z=0 (Io)
sst = np.loadtxt("/Users/ago/CompLifeSci/sst.dat") # sea surface temperature (T)
```

#### Calculate the derivative of the MLD as follows:

#### Create functions for the various processes as follows:

```
def uptake(n):
    u = n/(n+K_N)
    return u

def grazeP(p,d):
    g = (g_max*pP*p*p/(K_Z**2 + pP*p*p + pD*d*d))
    return g
```

#### Call these functions as appropriate when defining the differential equations:

```
def npzd(y, t):
    N, P, Z, D = y
    dPdt = (temp(sst[t])*uptake(N)*light(t,P))*P - mP*P - grazeP(P,D)*Z - mix(t)*P
    ...
    return dydt
```

# Further reading



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