Lagrangian or not Lagrangian?

Jean-Olivier Irisson, Michel de Lara, Serge Planes

ACI MOOREA EPHE - UMR CNRS 8046

> LAPCOD June 13, 2005

> > イロト イポト イヨト イヨト

3

Jean-Olivier Irisson, Michel de Lara, Serge Planes Lagrangian or not Lagrangian?

Introduction

The problem The solutions

2 Two models

A passive lagrangian model An active optimisation model

3 Advantages and drawbacks

Comparison Models developpement

< 🗇 🕨

* E > * E >

The problem The solutions



Jean-Olivier Irisson, Michel de Lara, Serge Planes

Lagrangian or not Lagrangian?

The problem The solutions

3

Ecological questions raised

• How many return?

The problem The solutions

-

Ecological questions raised

- How many return?
- Where do they go? Where do they come from? (dispersal vs. self recruitment)

The problem The solutions

Ecological questions raised

- How many return?
- Where do they go? Where do they come from? (dispersal vs. self recruitment)
- What are the parameters that control the trajectories?

Ecological questions raised

- How many return?
- Where do they go? Where do they come from? (dispersal vs. self recruitment)
- What are the parameters that control the trajectories?

Applications in:

Population genetics

< ロ > < 同 > < 三 > < 三 > 、

Ecological questions raised

- How many return?
- Where do they go? Where do they come from? (dispersal vs. self recruitment)
- What are the parameters that control the trajectories?

Applications in:

- Population genetics
- Conservation biology

Ecological questions raised

- How many return?
- Where do they go? Where do they come from? (dispersal vs. self recruitment)
- What are the parameters that control the trajectories?

Applications in:

- Population genetics
- Conservation biology
- Fisheries (professional and artisanal)

The problem The solutions

・ロン ・雪 と ・ ヨ と ・ ヨ と

3

Different approachs

direct fish sampling

The problem The solutions

イロト イポト イヨト イヨト

3

Different approachs

• direct fish sampling = costly and descriptive only

Jean-Olivier Irisson, Michel de Lara, Serge Planes Lagrangian or not Lagrangian?

The problem The solutions

イロン 不同 とくほう イロン

3

- direct fish sampling = costly and descriptive only
- genetics

The problem The solutions

イロト イポト イヨト イヨト

-

- direct fish sampling = costly and descriptive only
- genetics = not precise enough to estimate the fluxes of individuals

The problem The solutions

イロト イポト イヨト イヨト

-

- direct fish sampling = costly and descriptive only
- genetics = not precise enough to estimate the fluxes of individuals
- otolith rearing

The problem The solutions

・ロト ・回ト ・ヨト ・ヨト

-

- direct fish sampling = costly and descriptive only
- genetics = not precise enough to estimate the fluxes of individuals
- otolith rearing = lack of precision on the trajectory

The problem The solutions

イロン 不同 とくほう イロン

- 14

- direct fish sampling = costly and descriptive only
- genetics = not precise enough to estimate the fluxes of individuals
- otolith rearing = lack of precision on the trajectory
- modelling

The problem The solutions

(日)

- direct fish sampling = costly and descriptive only
- genetics = not precise enough to estimate the fluxes of individuals
- otolith rearing = lack of precision on the trajectory
- **modelling** = cheaper, precise if needed, inlights on the mechanisms but has to be validated

The problem The solutions

3

Different approachs

• **modelling** = cheaper, precise if needed, inlights on the mechanisms but has to be validated

A passive lagrangian model An active optimisation model

・ロン ・回 と ・ ヨ と ・ ヨ と …

3

Paris et al. in RSMAS, Miami

• Large scale model of the Wide Carabbean

A passive lagrangian model An active optimisation model

- Large scale model of the Wide Carabbean
- Coral reef habitat represented by 261 "polygons" of 100 $\rm km^2$ each

- Large scale model of the Wide Carabbean
- Coral reef habitat represented by 261 "polygons" of 100 $\rm km^2$ each
- Ocean circulation modelled with MICOM on grid-step of \approx 7km, for five years with realistic winds

- Large scale model of the Wide Carabbean
- Coral reef habitat represented by 261 "polygons" of 100 $\rm km^2$ each
- Ocean circulation modelled with MICOM on grid-step of \approx 7km, for five years with realistic winds
- Larvae released every month for pelagic larval duration of 30 days

- Large scale model of the Wide Carabbean
- Coral reef habitat represented by 261 "polygons" of 100 $\rm km^2$ each
- Ocean circulation modelled with MICOM on grid-step of \approx 7km, for five years with realistic winds
- Larvae released every month for pelagic larval duration of 30 days
- Larvae advected as passive particules using a Lagrangian model with Runge-Kutta integration and adaptative time-step

- Large scale model of the Wide Carabbean
- Coral reef habitat represented by 261 "polygons" of 100 $\rm km^2$ each
- Ocean circulation modelled with MICOM on grid-step of \approx 7km, for five years with realistic winds
- Larvae released every month for pelagic larval duration of 30 days
- Larvae advected as passive particules using a Lagrangian model with Runge-Kutta integration and adaptative time-step
- When encountering a polygon a larvae:

Paris et al. in RSMAS, Miami

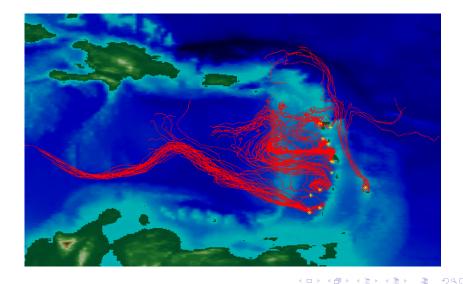
- Large scale model of the Wide Carabbean
- Coral reef habitat represented by 261 "polygons" of 100 $\rm km^2$ each
- Ocean circulation modelled with MICOM on grid-step of \approx 7km, for five years with realistic winds
- Larvae released every month for pelagic larval duration of 30 days
- Larvae advected as passive particules using a Lagrangian model with Runge-Kutta integration and adaptative time-step
- When encountering a polygon a larvae:

1 is retained if already "active" (=post-flexion)

- Large scale model of the Wide Carabbean
- Coral reef habitat represented by 261 "polygons" of 100 $\rm km^2$ each
- Ocean circulation modelled with MICOM on grid-step of \approx 7km, for five years with realistic winds
- Larvae released every month for pelagic larval duration of 30 days
- Larvae advected as passive particules using a Lagrangian model with Runge-Kutta integration and adaptative time-step
- When encountering a polygon a larvae:
 - **1** is retained if already "active" (=post-flexion)
 - 2 continues passive trajectory if not

A passive lagrangian model An active optimisation model

Precise trajectories

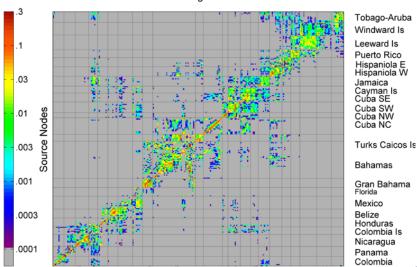


Jean-Olivier Irisson, Michel de Lara, Serge Planes

Lagrangian or not Lagrangian?

A passive lagrangian model An active optimisation model

Connectivity at the scale of the Carabbean



Receiving Nodes

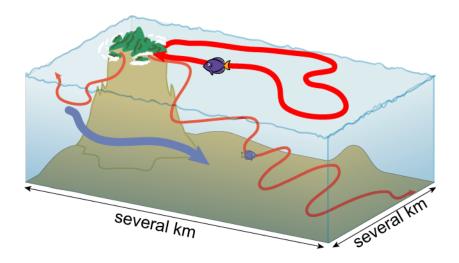
Jean-Olivier Irisson, Michel de Lara, Serge Planes

A passive lagrangian model An active optimisation model

・ロト ・回ト ・モト ・モト

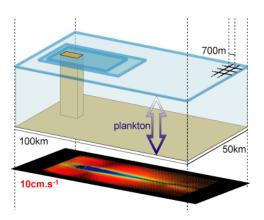
3

Our system



A passive lagrangian model An active optimisation model

Environment

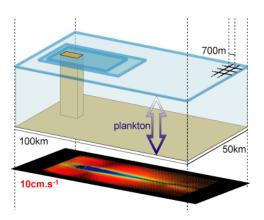


• Ocean circulation is constant and described by a 3D σ coordinate model (SYMPHONIE).

<ロ> <同> <同> < 回> < 回>

A passive lagrangian model An active optimisation model

Environment

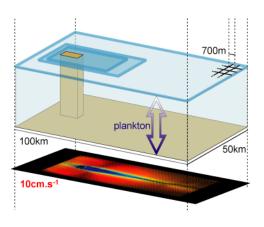


- Ocean circulation is constant and described by a 3D σ coordinate model (SYMPHONIE).
- Plankton and predators distribution (accounting for the *island mass effect*)

< ロ > < 同 > < 回 > < 回 >

A passive lagrangian model An active optimisation model

Environment



- Ocean circulation is constant and described by a 3D σ coordinate model (SYMPHONIE).
- Plankton and predators distribution (accounting for the *island mass effect*)
- Daily vertical migration of the plankton

イロト イポト イヨト イヨ

A passive lagrangian model An active optimisation model

Active larvae



• Pelagic larval duration of 20 days

<ロ> <同> <同> < 回> < 回>

Jean-Olivier Irisson, Michel de Lara, Serge Planes Lagrangian or not Lagrangian?

A passive lagrangian model An active optimisation model

Active larvae



- Pelagic larval duration of 20 days
- Three ontogenic stages. Swimming speeds = 3, 10, 20 cm.s⁻¹; Energetic needs = vitelus then one unit per unit time.

< ロ > < 同 > < 回 > < 回 >

A passive lagrangian model An active optimisation model

Active larvae

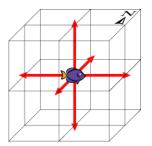


- Pelagic larval duration of 20 days
- Three ontogenic stages. Swimming speeds = 3, 10, 20 cm.s⁻¹; Energetic needs = vitelus then one unit per unit time.
- Energy budget ⇒ more than one day without eating means death.

イロト イポト イヨト イヨ

Trajectories controlled by active larvae

• Simple advection scheme (⇔"first order" Lagrangian).

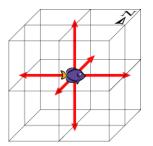


A passive lagrangian model An active optimisation model

・ロン ・四 と ・ ヨ と ・ ヨ と ・

Trajectories controlled by active larvae

- Simple advection scheme (⇔ "first order" Lagrangian).
- Activity of the larvae: two behavioural alternatives at each time step (6h):

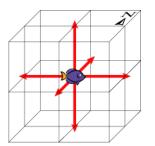


・ 同 ト ・ ヨ ト ・ ヨ ト

Trajectories controlled by active larvae

- Simple advection scheme (⇔ "first order" Lagrangian).
- Activity of the larvae: two behavioural alternatives at each time step (6h):

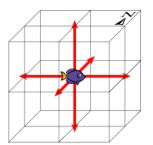
1 feeding and gaining energy but swimming randomly



・ 同 ト ・ ヨ ト ・ ヨ ト

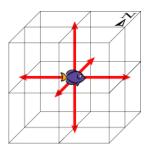
Trajectories controlled by active larvae

- Simple advection scheme (⇔ "first order" Lagrangian).
- Activity of the larvae: two behavioural alternatives at each time step (6h):
 - 1 feeding and gaining energy but swimming randomly
 - 2 swimming in a precise direction but loosing energy



Trajectories controlled by active larvae

- Simple advection scheme (⇔ "first order" Lagrangian).
- Activity of the larvae: two behavioural alternatives at each time step (6h):
 - 1 feeding and gaining energy but swimming randomly
 - 2 swimming in a precise direction but loosing energy

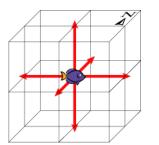




・ロト ・回ト ・ヨト・

Trajectories controlled by active larvae

- Simple advection scheme (⇔ "first order" Lagrangian).
- Activity of the larvae: two behavioural alternatives at each time step (6h):
 - 1 feeding and gaining energy but swimming randomly
 - 2 swimming in a precise direction but loosing energy



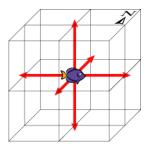
$\Rightarrow \textbf{Choice}$

• The choice is made depending on environnemental factors and on its state

・ロト ・雪 ト ・ ヨ ト・

Trajectories controlled by active larvae

- Simple advection scheme (⇔ "first order" Lagrangian).
- Activity of the larvae: two behavioural alternatives at each time step (6h):
 - 1 feeding and gaining energy but swimming randomly
 - 2 swimming in a precise direction but loosing energy



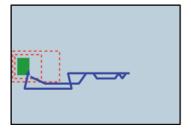
$\Rightarrow \textbf{Choice}$

- The choice is made depending on environnemental factors and on its state
- Choice criteria: "optimal" choice

・ロン ・四 と ・ ヨ と ・ ヨ と ・

A passive lagrangian model An active optimisation model

Results



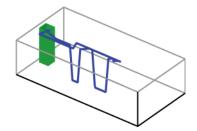


Image: Image:

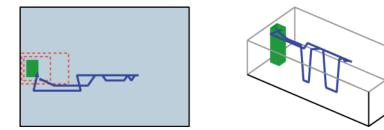
A B > A B >

Larvae following the optimal trajectories avoid high predation areas, maximize the probability of finging food and swim against currents when it's the most efficient

A passive lagrangian model An active optimisation model

<ロト <同ト < 国ト < ヨト

Results



Larvae following the optimal trajectories avoid high predation areas, maximize the probability of finging food and swim against currents when it's the most efficient

captures **fine scale** properties of the trajectory, related to the **behavioural** abilities of the larvae and far from Lagrangian trajectories

Comparison Models developpement

Some differences

Lagrangian

Optimisation

・ロン ・回 と ・ ヨ と ・ ヨ と …

3

Jean-Olivier Irisson, Michel de Lara, Serge Planes Lagrangian or not Lagrangian?

Comparison Models developpement

Some differences

Lagrangian

• Large scale (space vectorized)

Optimisation

 Small scale (space discretized and dimension is costly in the optimization process)

-

Comparison Models developpement

Some differences

Lagrangian

- Large scale (space vectorized)
- Focus on the oceanography rather than on the biology

Optimisation

- Small scale (space discretized and dimension is costly in the optimization process)
- Focus on the biology rather than on the oceanography

< 17 ▶

A B > A B >

Comparison Models developpement

Some differences

Lagrangian

- Large scale (space vectorized)
- Focus on the oceanography rather than on the biology
- Precision of the prediction
 ⇒ direct validation and
 application is possible
 (tracking of larvae schools,
 genetic connectivity...)
- Predicts the influence of some oceanographic regimes

Optimisation

- Small scale (space discretized and dimension is costly in the optimization process)
- Focus on the biology rather than on the oceanography

< ロ > < 同 > < 回 > < 回 > < 回 > <

 Model aimed at understanding the processes (extensive sensitivity analysis)

Comparison Models developpement

3

Why not both?

- The optimization process requires a discretized state \Rightarrow prevents large scale

イロト イポト イヨト イヨト

- The optimization process requires a discretized state \Rightarrow prevents large scale
- Transitions described by transition probabilities stored in huge matrices wich have to be loaded in memory ⇒ there cannot be too many

・ロン ・部 と ・ ヨ と ・ ヨ と …

- The optimization process requires a discretized state \Rightarrow prevents large scale
- Transitions described by transition probabilities stored in huge matrices wich have to be loaded in memory \Rightarrow there cannot be too many
- To choose the right decision, all have to be computed \Rightarrow prevents a precise Lagrangian advection scheme

Comparison Models developpement

-

< 17 ▶

Future developments

Lagrangian

 Better oceanographic model with vertical resolution and random vertical migration

Comparison Models developpement

Future developments

Lagrangian

 Better oceanographic model with vertical resolution and random vertical migration

Optimisation

• Re-writing of the code and parallelisation to gain dimension

Image: Image:

A B + A B +

Comparison Models developpement

Future developments

Lagrangian

 Better oceanographic model with vertical resolution and random vertical migration

Optimisation

- Re-writing of the code and parallelisation to gain dimension
- New optimization method using genetic algorithm

A B + A B +

・ロト ・回ト ・ヨト ・ヨト

3

Thank you for your attention