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ReVa



Beneficial Catastrophes: Leveraging Abiotic Constraints through Environment-Driven Evolutionary Selection

Kevin GODIN-DUBOIS

On-demand pre-recorded presentation
for IEEE 2020 ALife Symposium



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ReVa

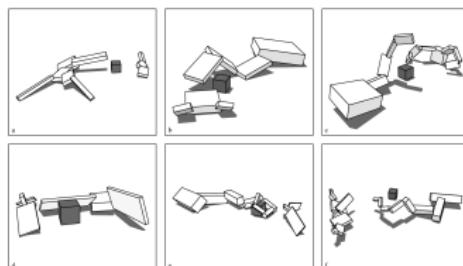


Beneficial Catastrophes: Leveraging Abiotic Constraints through Environment-Driven Evolutionary Selection

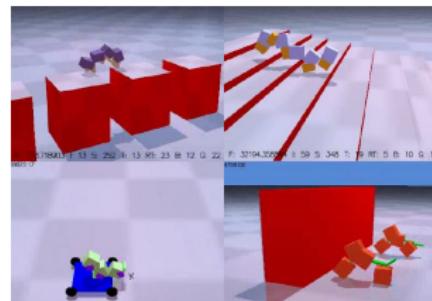
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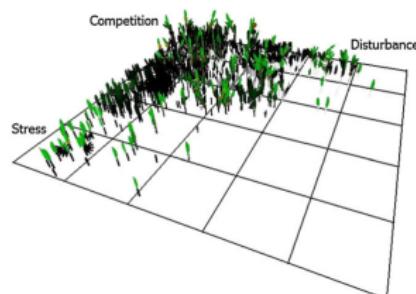
Context



Sims (1994)



Lassabe et al. (2007)

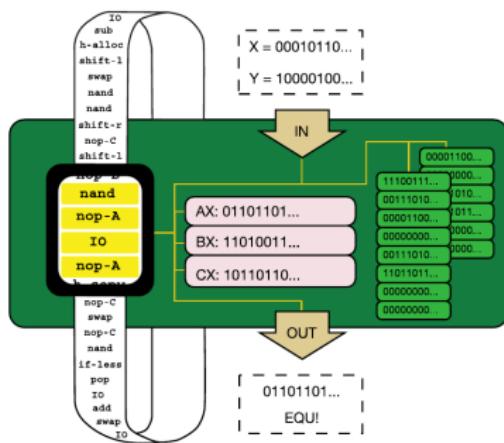


Bornhofen et al. (2011)

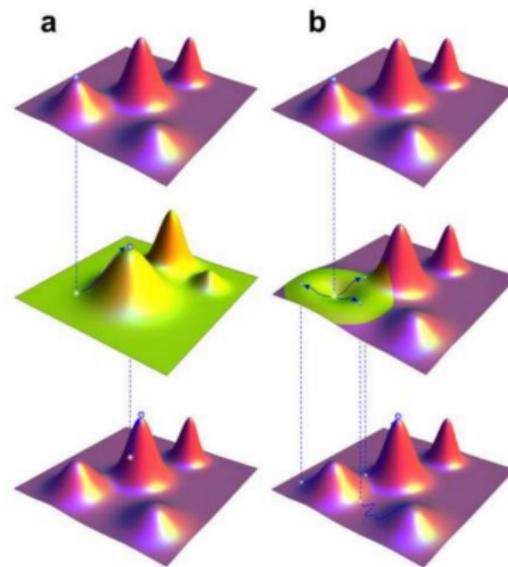


Eloy et al. (2017)

Context



Lenski et al. (2003)



Nahum et al. (2017)

EDEnS

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Promoting robustness

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Insights

○○○○○

Conclusion

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EDEnS

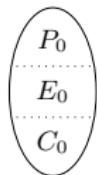
EDEnS

Environment-Driven Evolutionary Selection

- Parallel exploration of alternative “realities”
- Guided by a set of fitness functions
- Evaluating populations
- Controlling *indirectly* through environments

EDEnS

Main steps



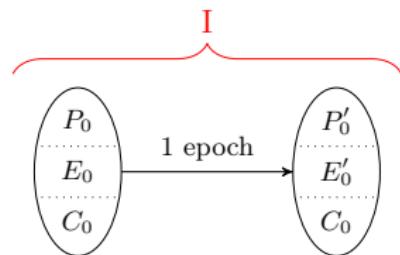
P/P' : pop. before/after evolution

E/E' : env. before/after variations

C : env. controller

EDEnS

Main steps



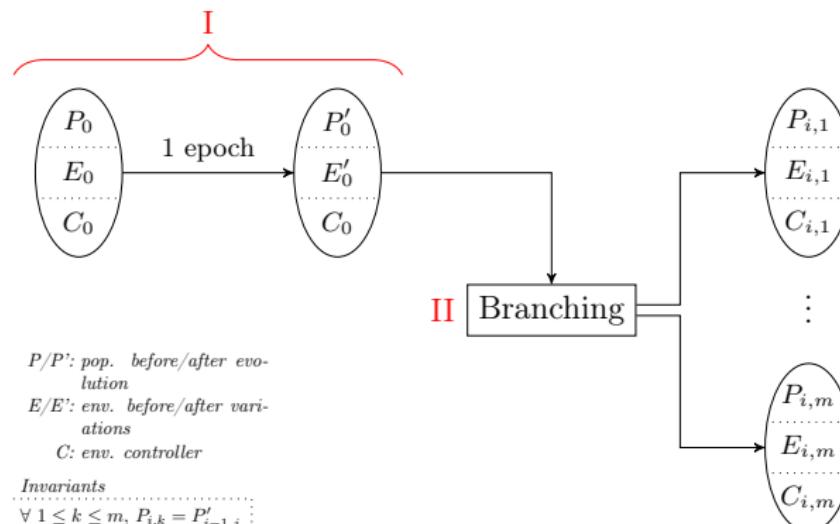
P/P' : pop. before/after evolution

E/E' : env. before/after variations

C : env. controller

EDEnS

Main steps



P/P' : pop. before/after evolution

E/E' : env. before/after variations

C : env. controller

Invariants

$$\forall 1 \leq k \leq m, P_{i,k} = P'_{i-1,j}$$

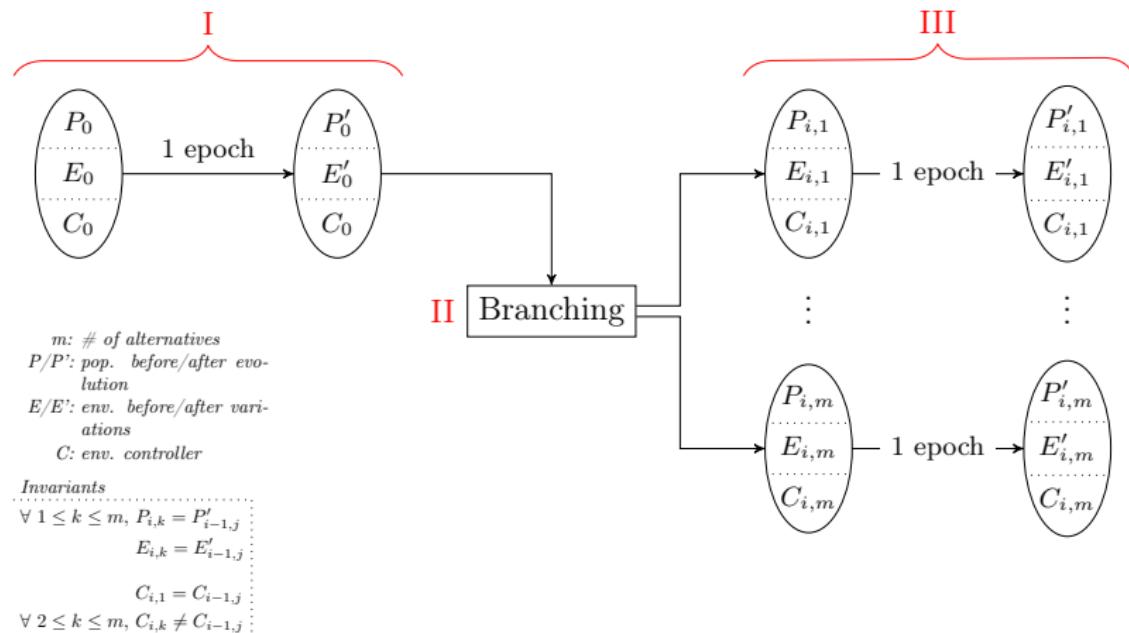
$$E_{i,k} = E'_{i-1,j}$$

$$C_{i,1} = C_{i-1,j}$$

$$\forall 2 \leq k \leq m, C_{i,k} \neq C_{i-1,j}$$

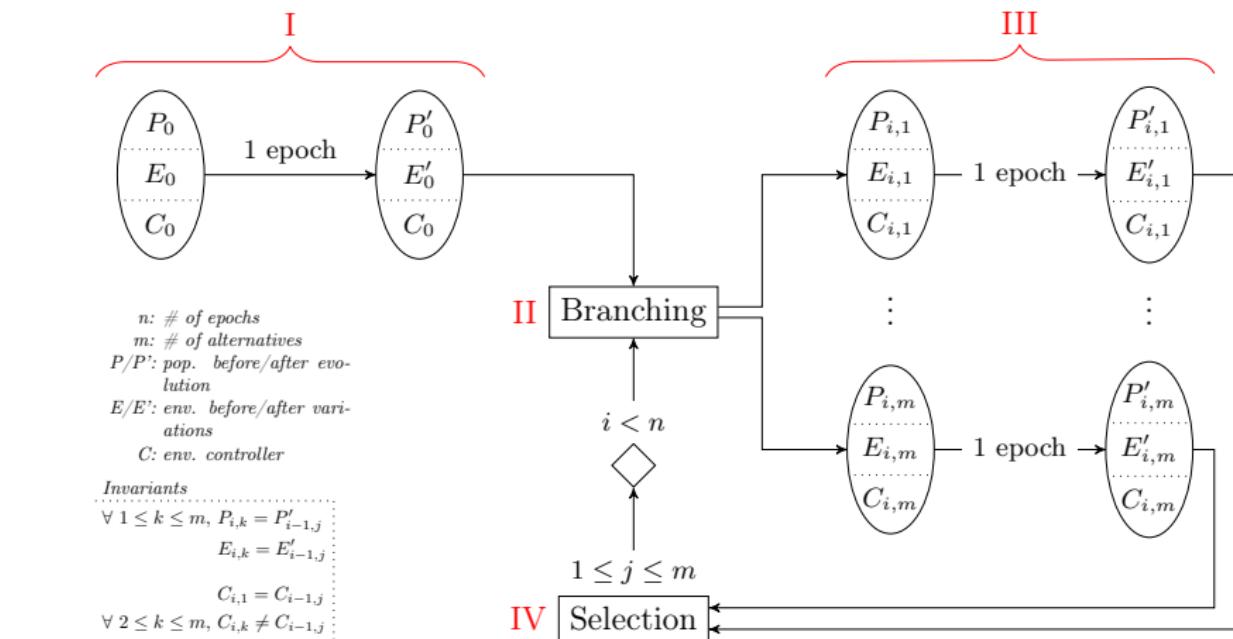
EDEnS

Main steps



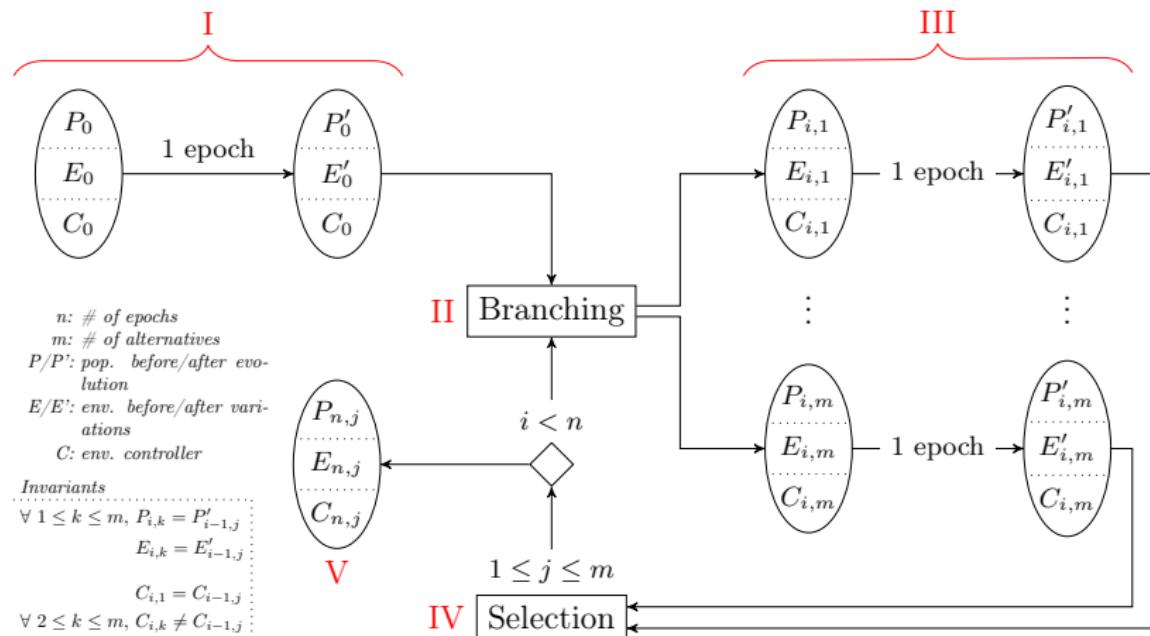
EDEnS

Main steps



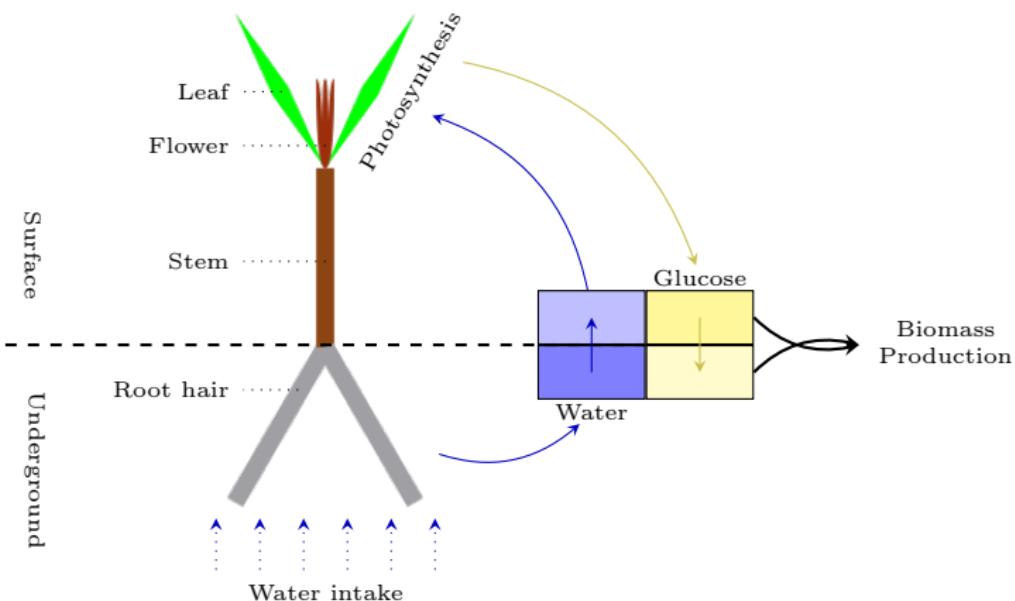
EDEnS

Main steps



Implementation

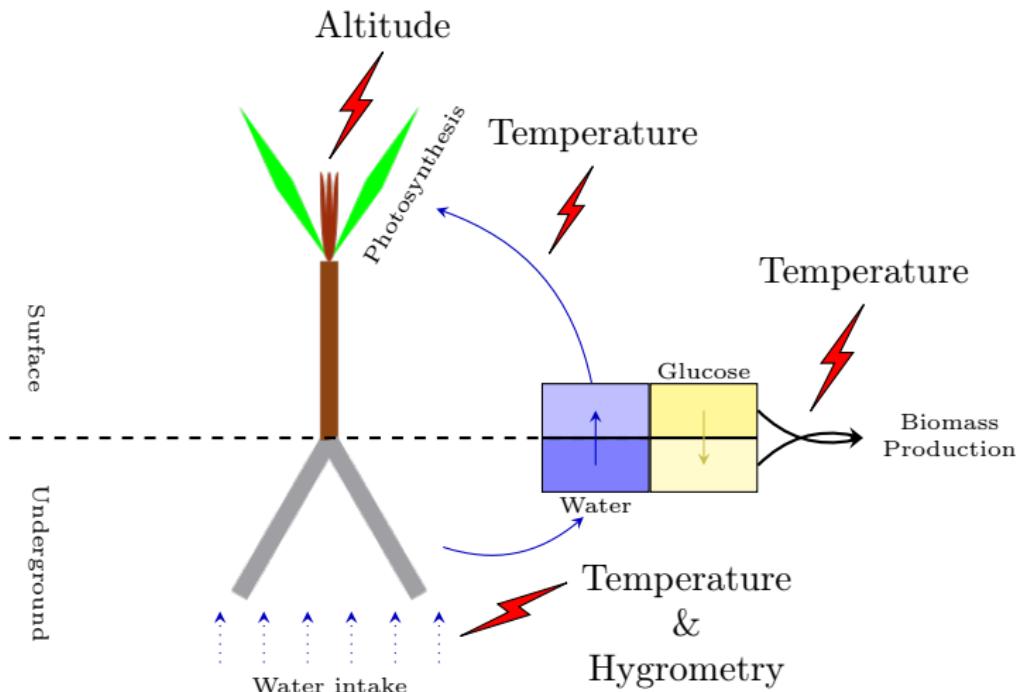
Plants



Full implementation details can be found in [3]

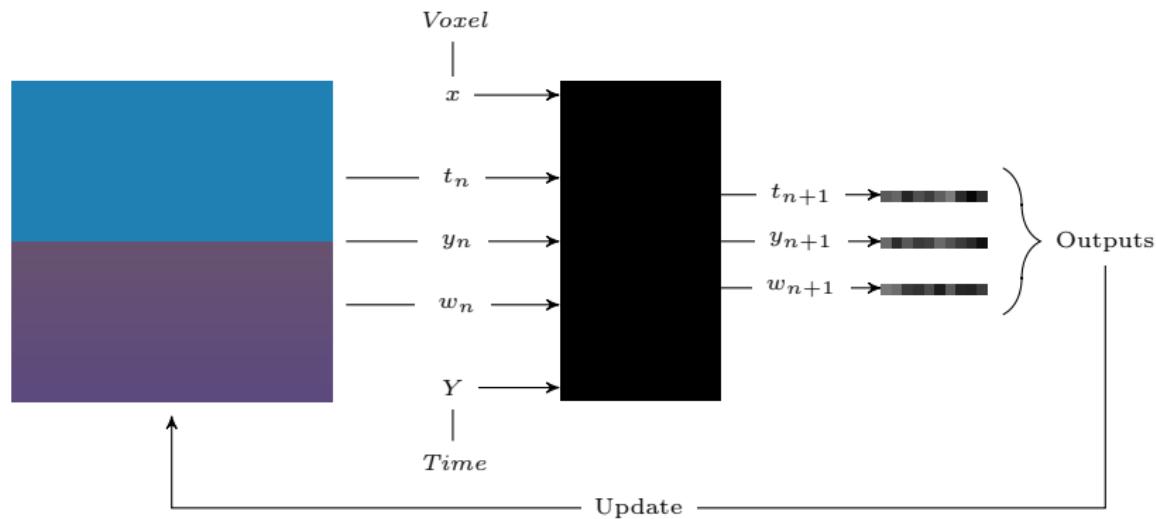
Implementation

Plants

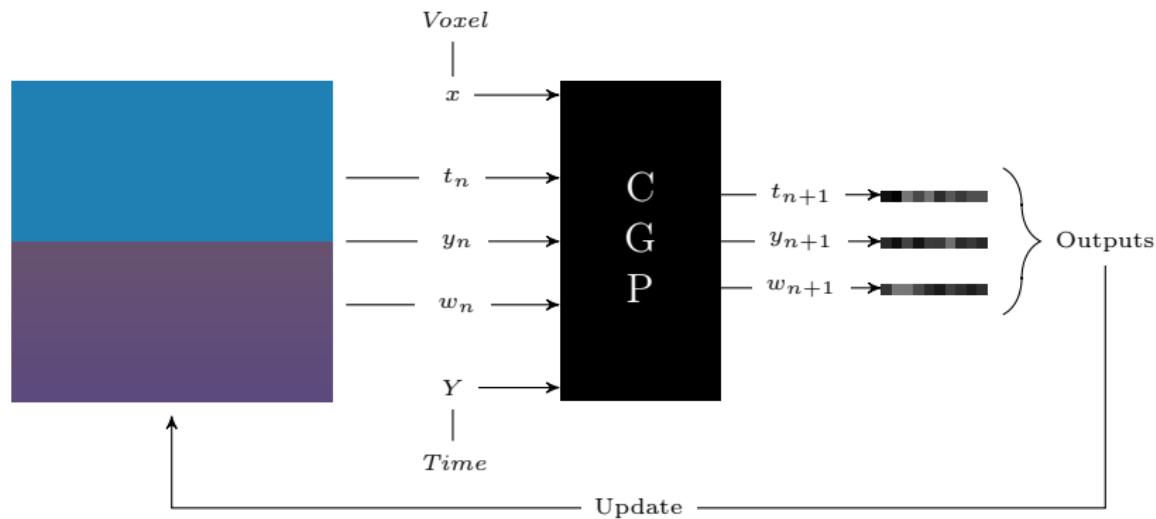


Full implementation details can be found in [3]

Implementation Environment



Implementation Environment



Promoting robustness

Promoting robustness

Evolutionary protocol

Fitness functions		Parameters	
C_P	Population size	n = 250	epochs
C_T	Evaluation time	e = 4	years per epoch
F_S	Speciation patterns	m = 10	branching factor
F_D	Genetic diversity		

Two groups:

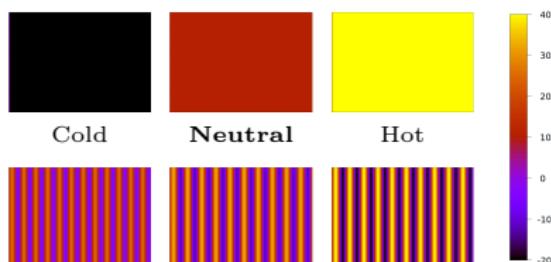
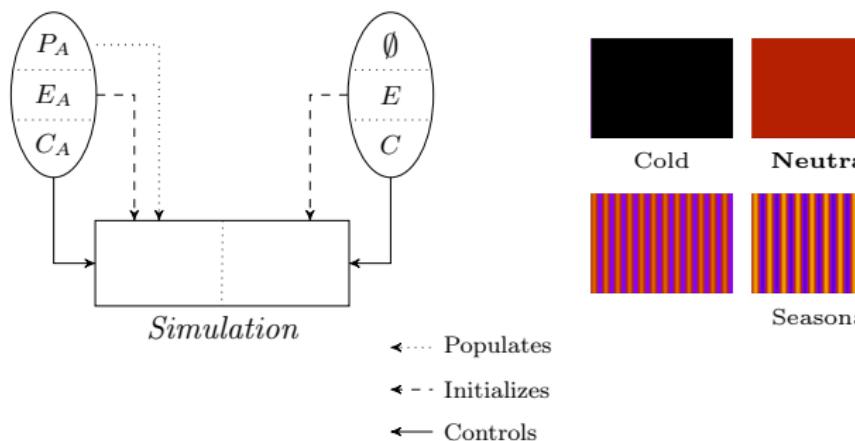
Evolved (e) uses EDEnS, 10 replicates

Control (c) hospitable conditions, 5 replicates

Promoting robustness

PvE

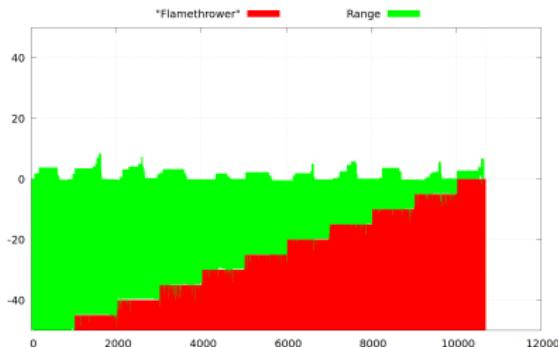
Hand-crafted right-hand side conditions



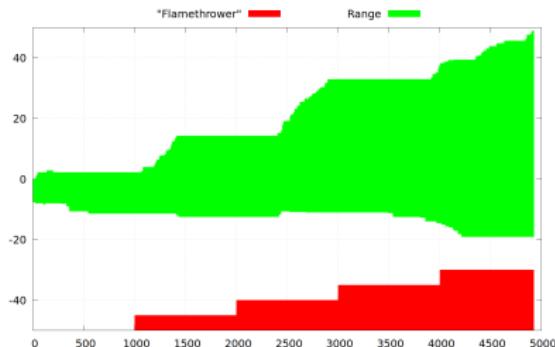
Promoting robustness

PvE

Score is the distance “traveled”



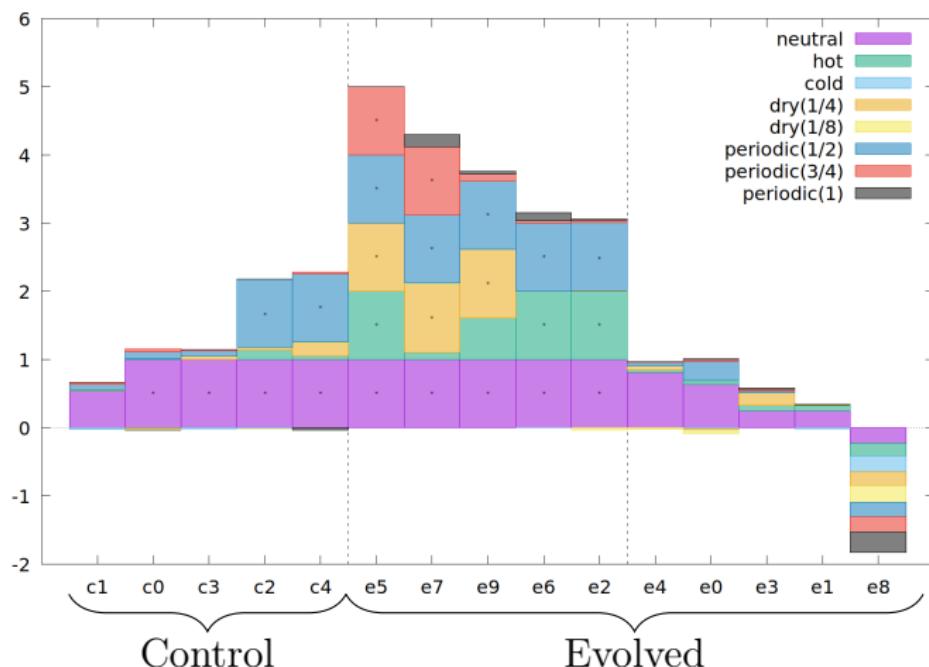
Poor performance (score = 0)



Goal reached (score = 1)

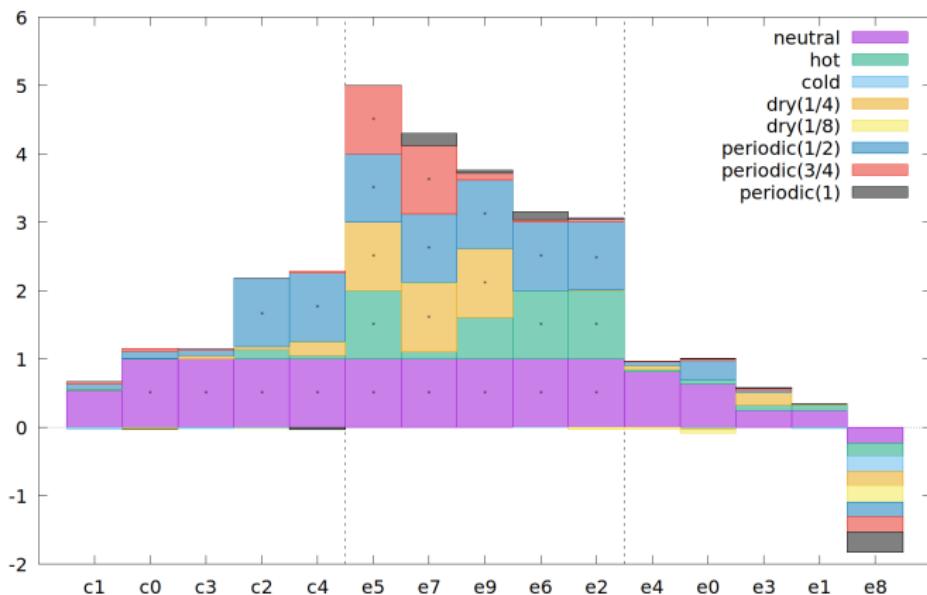
Promoting robustness

PvE



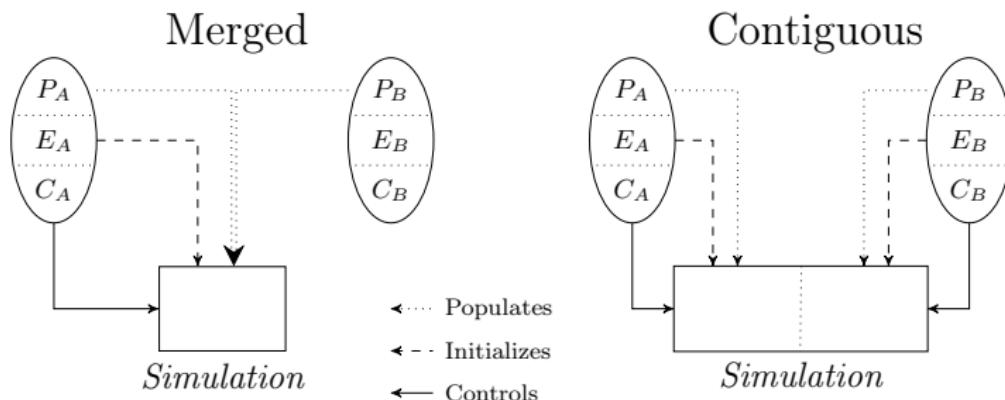
Promoting robustness

PvE

Evolved group (*e*) out-performs Control¹¹Mann-Whitney, p-value < 0.05, outlier *e*₈ not considered

Promoting robustness

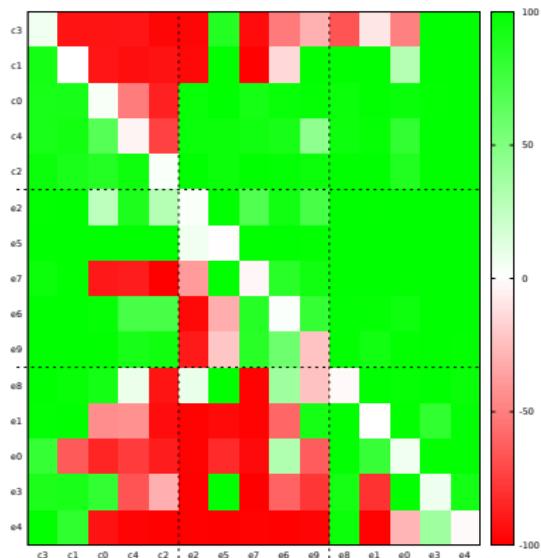
PvP



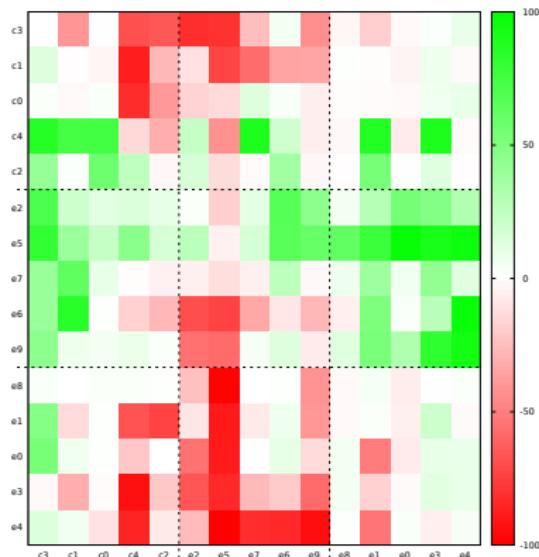
Promoting robustness

PvP

Merged

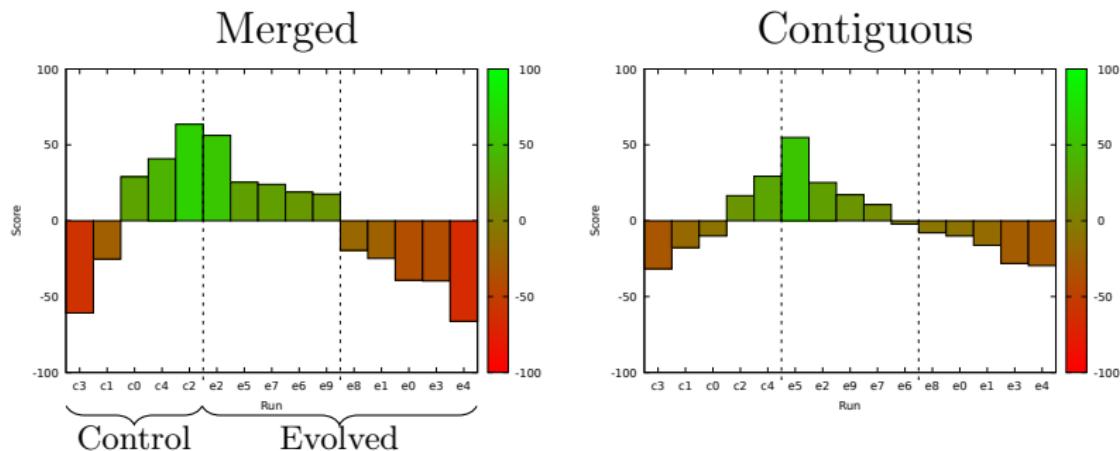


Contiguous



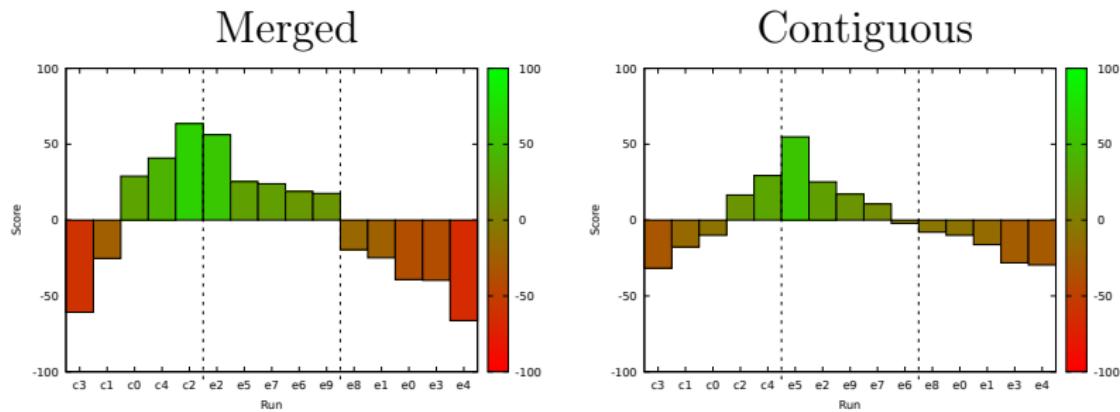
Promoting robustness

PvP



Promoting robustness

PvP



Statistical differences²:

- Merged: e out-performed by c
- Contiguous: not significant

²Wilcoxon test, p-value < 0.05

EDEnS
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Promoting robustness
○○○○

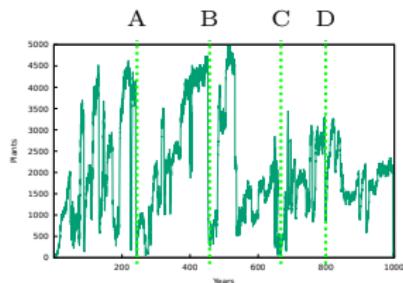
Insights
●○○○○

Conclusion
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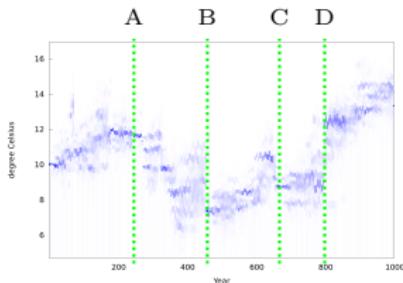
Insights

Individual strategies

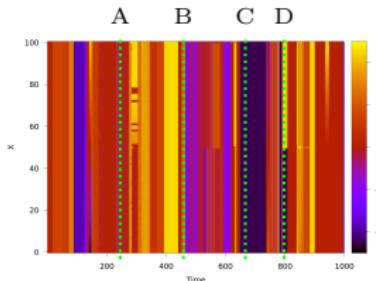
e_5 : PvE and Contiguous PvP champion



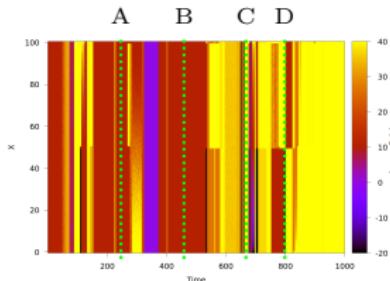
Variations of population size



Optimal temperature (gene)



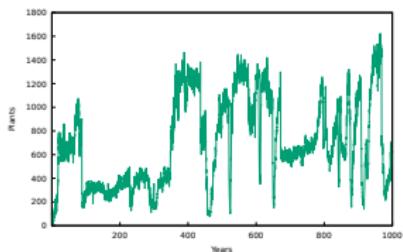
Environmental topography



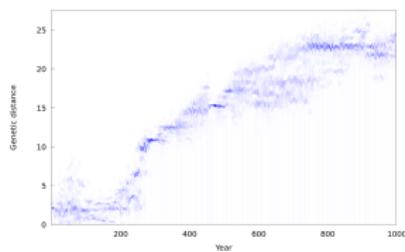
Environmental temperature

Individual strategies

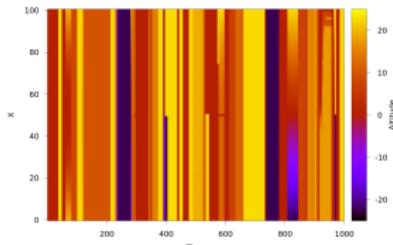
e_4 : worst member of evolved group



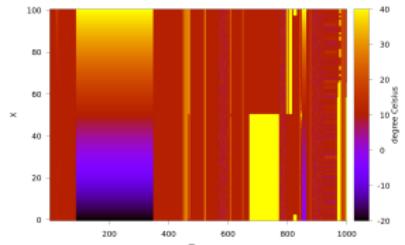
Variations of population size



Genetic distance (gene)



Environmental topography



Environmental temperature

Beneficial catastrophes

$$d_e = P_l(4(e - \nu)) - P_l(4(e + \nu))$$

$$d_e^- = \begin{cases} d_e & \text{if } d_e < 0 \\ 0 & \text{otherwise} \end{cases}$$

$$P_r = \sum_{e=1}^{249} \frac{|d_e|}{249}$$

$$P_r^- = - \sum_{e=1}^{249} \frac{d_e^-}{249}$$

$$\bar{P}_r = \frac{P_r}{\bar{P}_l} \quad \bar{P}_r^- = \frac{P_r^-}{\bar{P}_l}$$

with

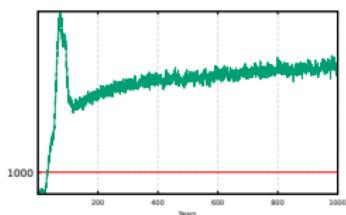
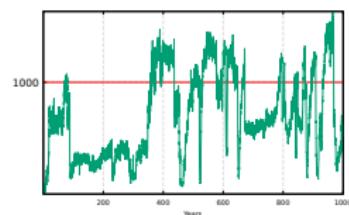
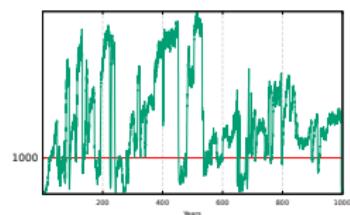
$P_l(t)$: Population size at time t

e : Time in years

ν : Integration window (0.25)

\bar{P}_l : Average population size

Beneficial catastrophes

 c_0  e_4  e_5 

$\bar{P}_r = .02 \quad \bar{P}_r^- = .02$

$\bar{P}_r = .07 \quad \bar{P}_r^- = .08$

$\bar{P}_r = .11 \quad \bar{P}_r^- = .16$

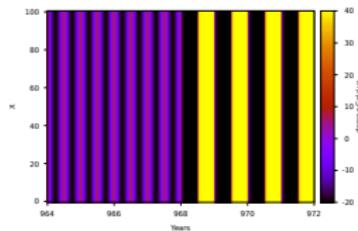
Beneficial catastrophes

Variable	<i>c & e</i>			<i>e</i> only		
	PvE	PvP		PvE	PvP	
		M	C		M	C
\bar{P}_l	.696	.589	.421	.697	.648	.697
P_r	.757	.625	.682	.685	.697	.733
P_r^-	.793	.589	.679	.770	.661	.733
\bar{P}_r	.564	.125	.511	.891	.758	.794
\bar{P}_r^-	.529	.193	.571	.782	.927	.927

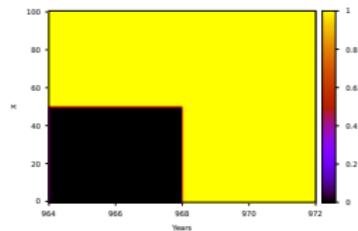
Positive correlations between scores and perturbation intensity³

³Spearman, grayed-out values are not significant under p-value < 0.05

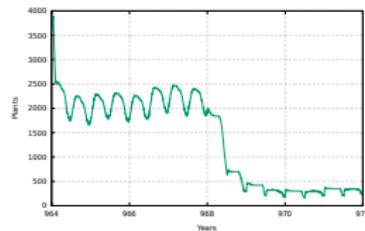
Resilience & Creativity



Temperature



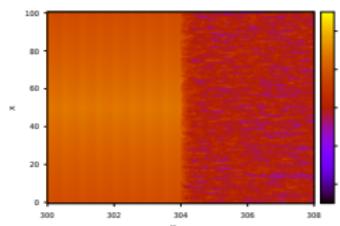
Hygrometry



Population

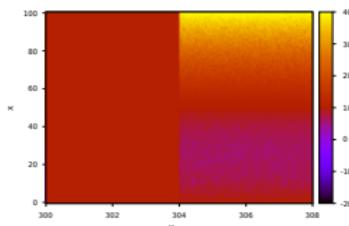
Resilience & Creativity

$$T = \text{rand}$$



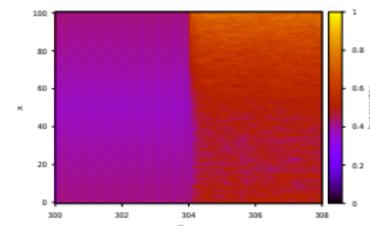
Topography

$$H(X) = X(\text{rand} < X)$$

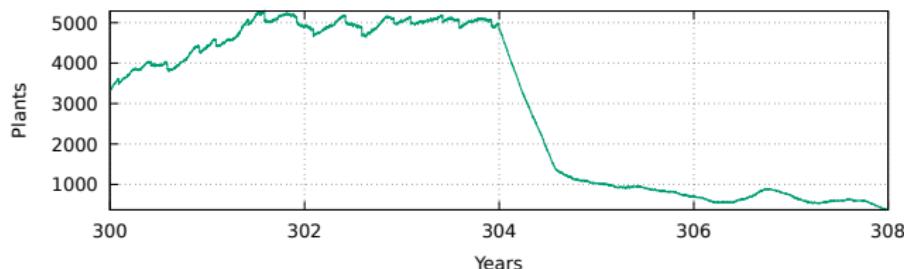


Temperature

$$W(T, H) = H - T$$



Hygrometry



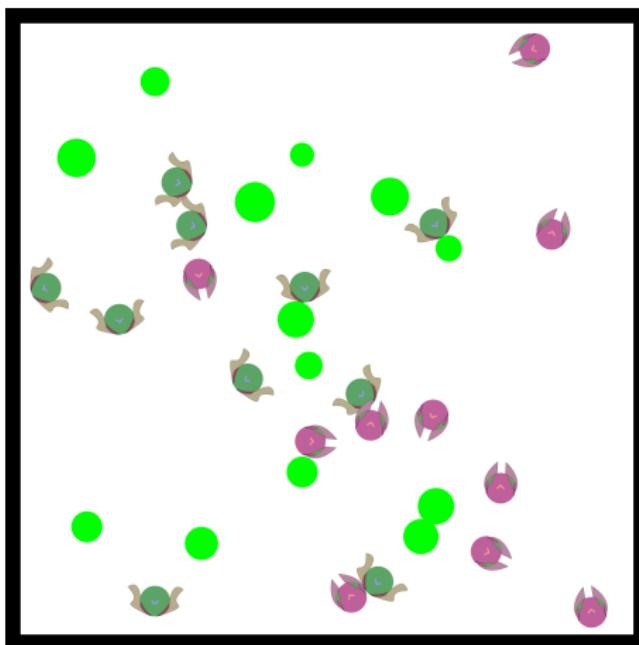
Population

Conclusion

Conclusion

- EDEnS: Environment-Driven Evolutionary Selection
- Application to abiotic robustness of virtual plants
- Showcased benefits of catastrophic trimming
- Generated out-of-the-box sets of constraints

Splinoids



Annexes

- References

References

-  S. Bornhofen, S. Barot, and C. Lattaud. "The evolution of CSR life-history strategies in a plant model with explicit physiology and architecture". In: *Ecological Modelling* 222.1 (Jan. 2011), pp. 1–10.
-  Christophe Eloy et al. "Wind loads and competition for light sculpt trees into self-similar structures". In: *Nature Communications* 8.1 (Dec. 2017), p. 1014.
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-  Richard E. Lenski et al. "The evolutionary origin of complex features". In: *Nature* 423.6936 (May 2003), pp. 139–144.
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