

Genetic Improvement of Shoreline Evolution Forecasting Models

Mahmoud Al Najar, Rafael Almar, Erwin W. J. Bergsma, Jean-Marc Delvit, Dennis G. Wilson



Coastal Zones

- High societal, ecological and economic importance
- Under increasing anthropogenic and climate pressures



Figure: "Schematic representation of the coastal zone, hazards, and metocean variables that are relevant for coastal marine hazards and their monitoring". Source: [1]

- Goal: improved forecasts of wave-driven shoreline change

[1] Melet, A. et al. (2020). Earth observations for monitoring marine coastal hazards and their drivers. *Surveys in Geophysics*, 41(6), 1489-1534.

Wave-driven shoreline change

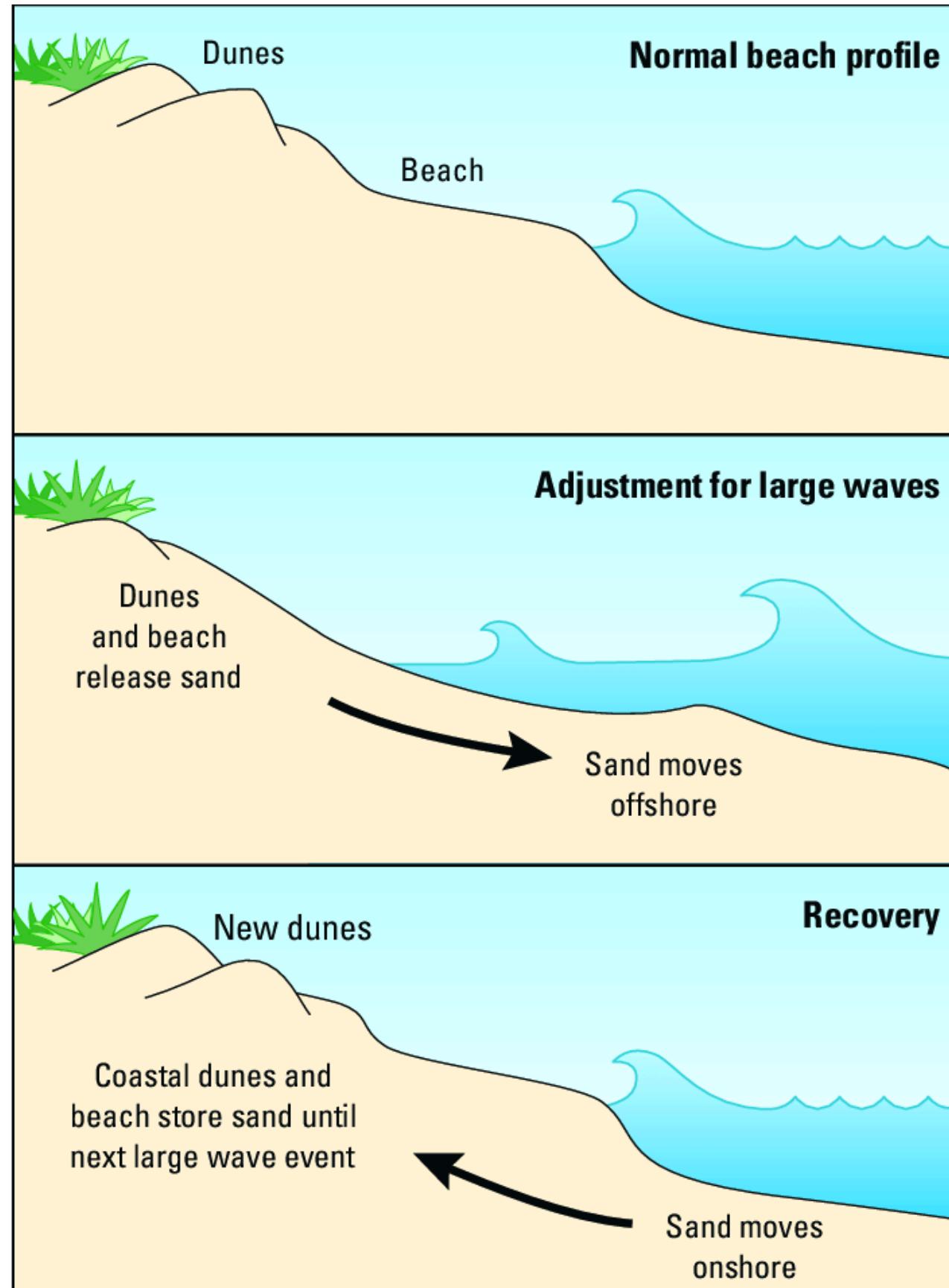


Figure: Seasonal beach-profile adjustments induced by seasonal swell variations and resulting cross-shore sediment transport. Source: [2]

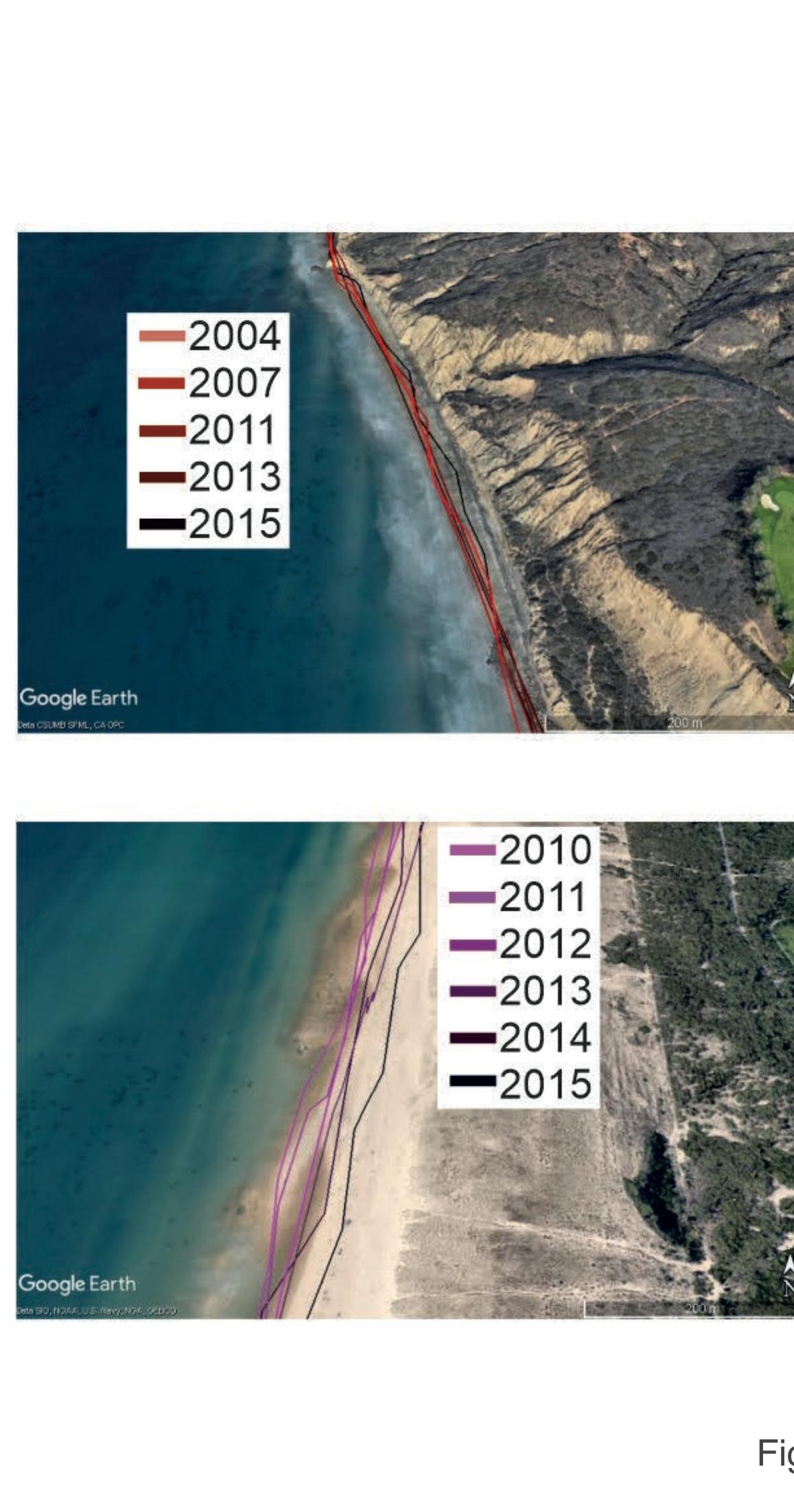
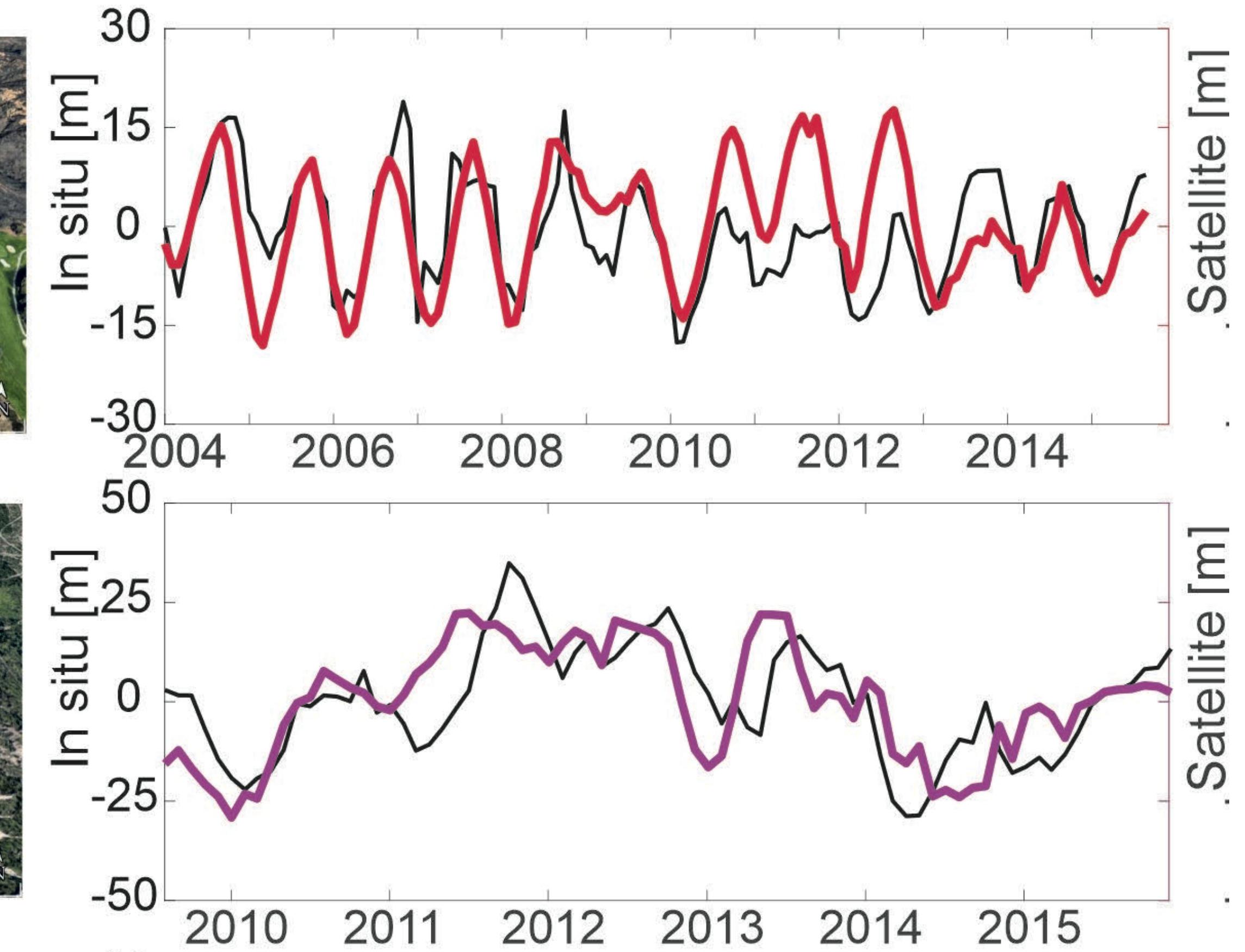


Figure: Illustration of shoreline variability. Source: [3]



[2] Fletcher, C. H. et al. (2012). National assessment of shoreline change: Historical shoreline change in the Hawaiian Islands.

[3] Graffin, M. et al. (2022). El Niño controls the evolution of shorelines worldwide. (In review, preprint available online)

Genetic Improvement of Shoreline Equilibrium Models

Outline

- 1) Coastal zones & shoreline change**
- 2) Study zone & dataset**
- 3) Existing methods**
- 4) Proposed CGP/GI formulation**
- 5) Experiments & results**
- 6) Future directions & conclusions**

Study Zone: Tairua, NZ

- Data was obtained from the “Shoreshop” competition [4]
- ~15 years of (open) data:
 - Daily video-derived shoreline positions
 - Hourly wave heights (H_s), periods (T_p) and directions (Dir)

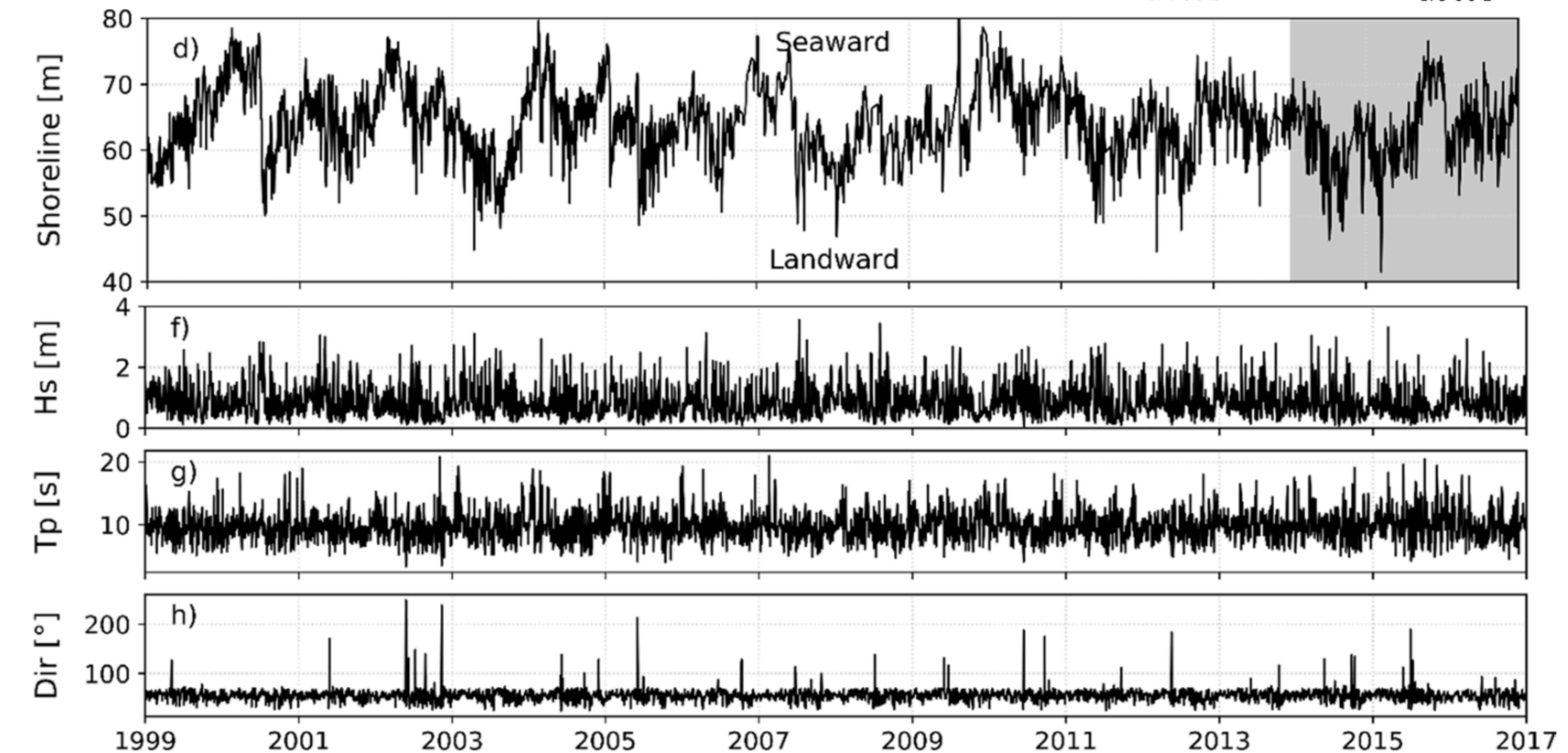


Figure: Tairua dataset. Source: [4]

Existing methods for shoreline forecasting

- Simulation-based methods
- Data-driven models
- Hybrid models
 - Embedded physical knowledge
 - Data-driven calibration

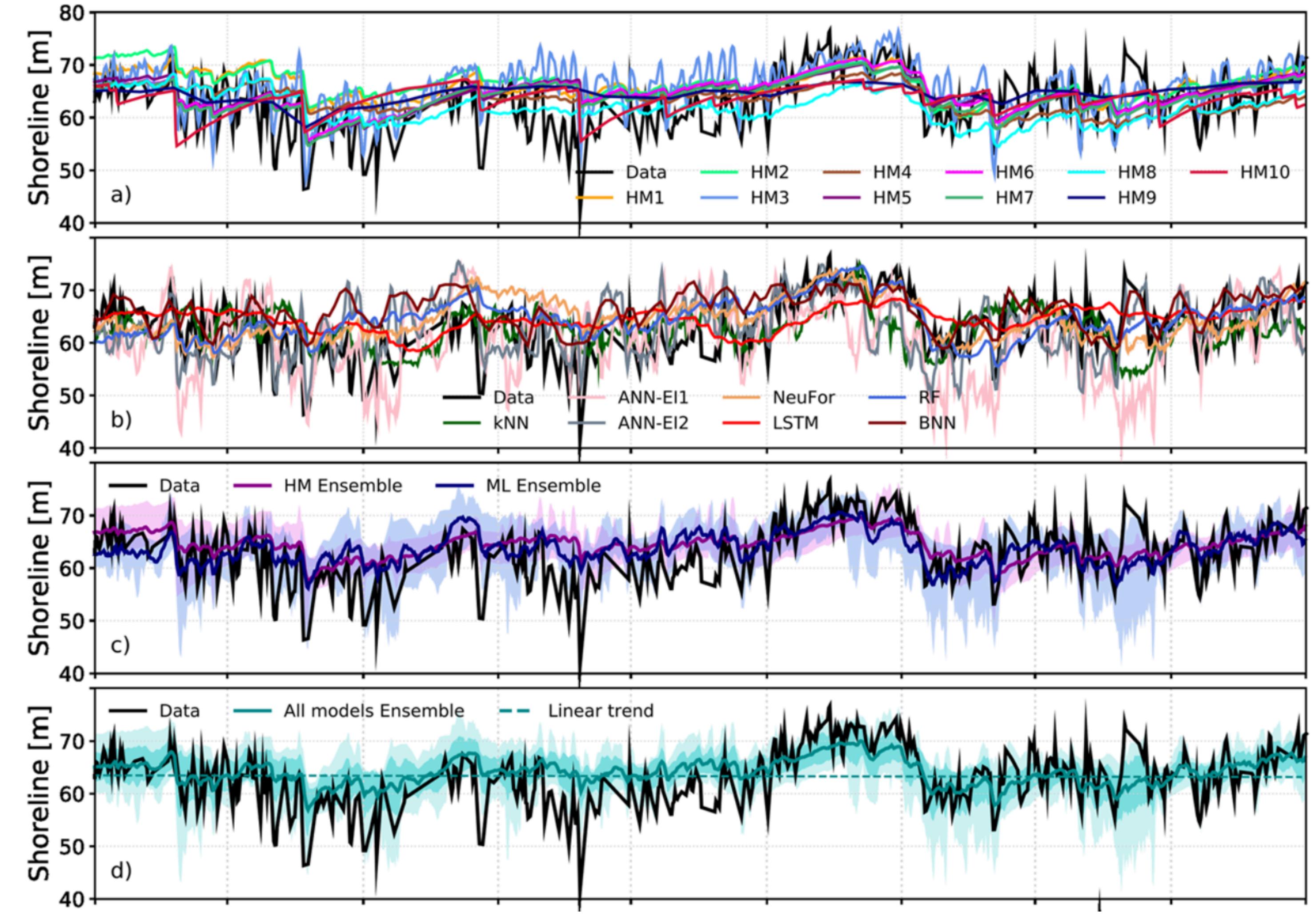


Figure: Performance comparison of hybrid and data-driven models included in the Shoreshop competition.

Source: [4]

ShoreFor

- ShoreFor: hybrid model [4,5] based on beach memory and equilibrium
- Variants proposed accounting for additional coastal processes and time scales

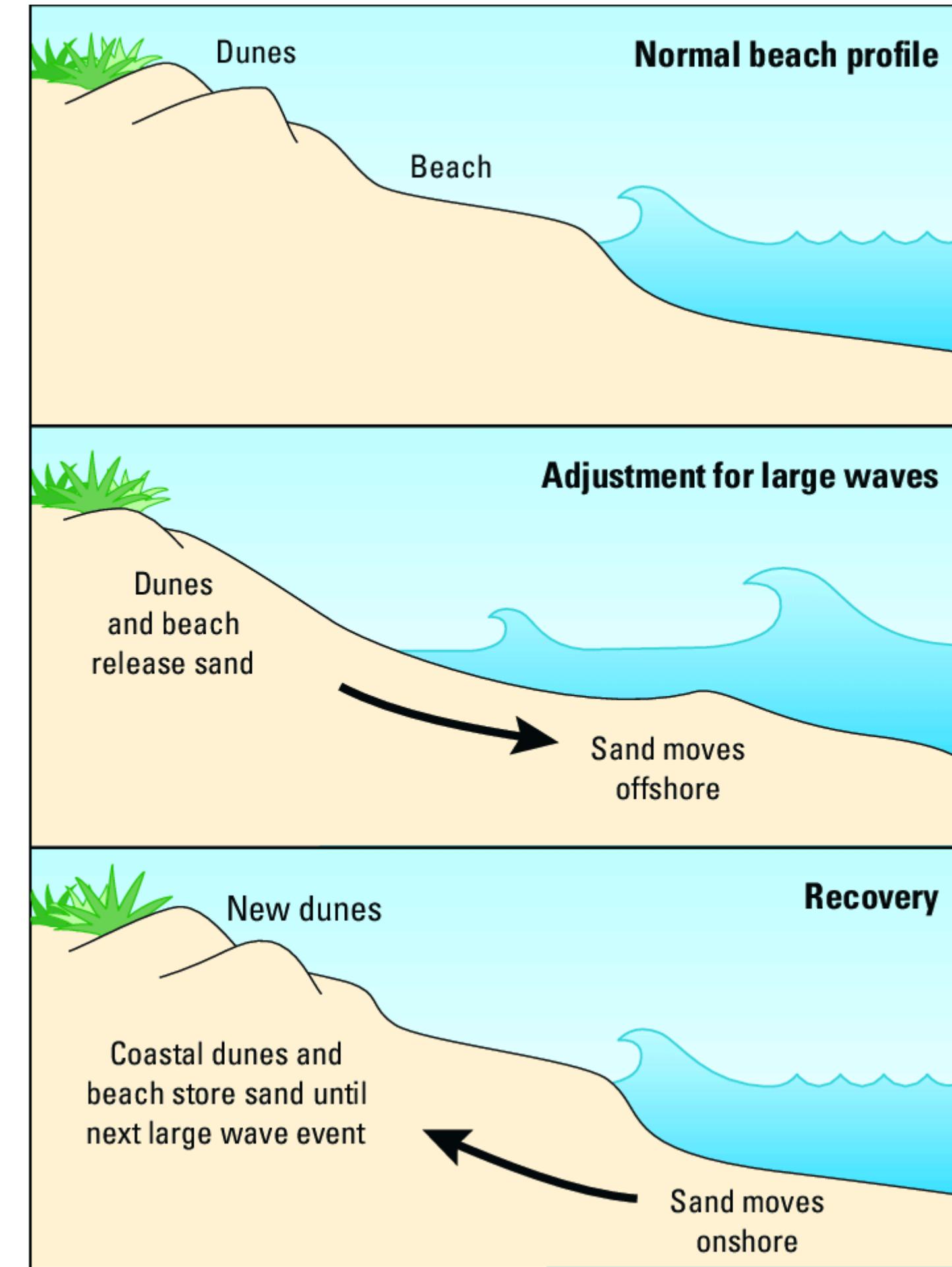


Figure: Seasonal beach-profile adjustments induced by seasonal swell variations and resulting cross-shore sediment transport. Source: [2]

ShoreFor
$\frac{dx}{dt} = c(F^+ + rF^-) + b$
$\Omega_{eq} = \frac{\sum_{i=1}^{2\phi} \Omega_i 10^{-i/\phi}}{\sum_{i=1}^{2\phi} 10^{-i/\phi}}$
$r = \left \frac{\sum_{i=0}^N \langle F_i^+ \rangle}{\sum_{i=0}^N \langle F_i^- \rangle} \right $
$F = P^{0.5} \frac{\Delta\Omega}{\sigma_{\Delta\Omega}}$
$\Omega = \frac{H_{s,b}}{wT_p}$

Table: ShoreFor system of equations

[4] Davidson, M. A., Splinter, K. D. & Turner, I. L. A simple equilibrium model for predicting shoreline change. *Coast. Eng.* 73, 191–202 (2013).

[5] Splinter, K. D. et al. A generalized equilibrium model for predicting daily to interannual shoreline response. *J. Geophys. Res. F Earth Surf.* 119, 1–23 (2014)

Cartesian Genetic Programming

- Relatively low-data regime
- “White-box” approach

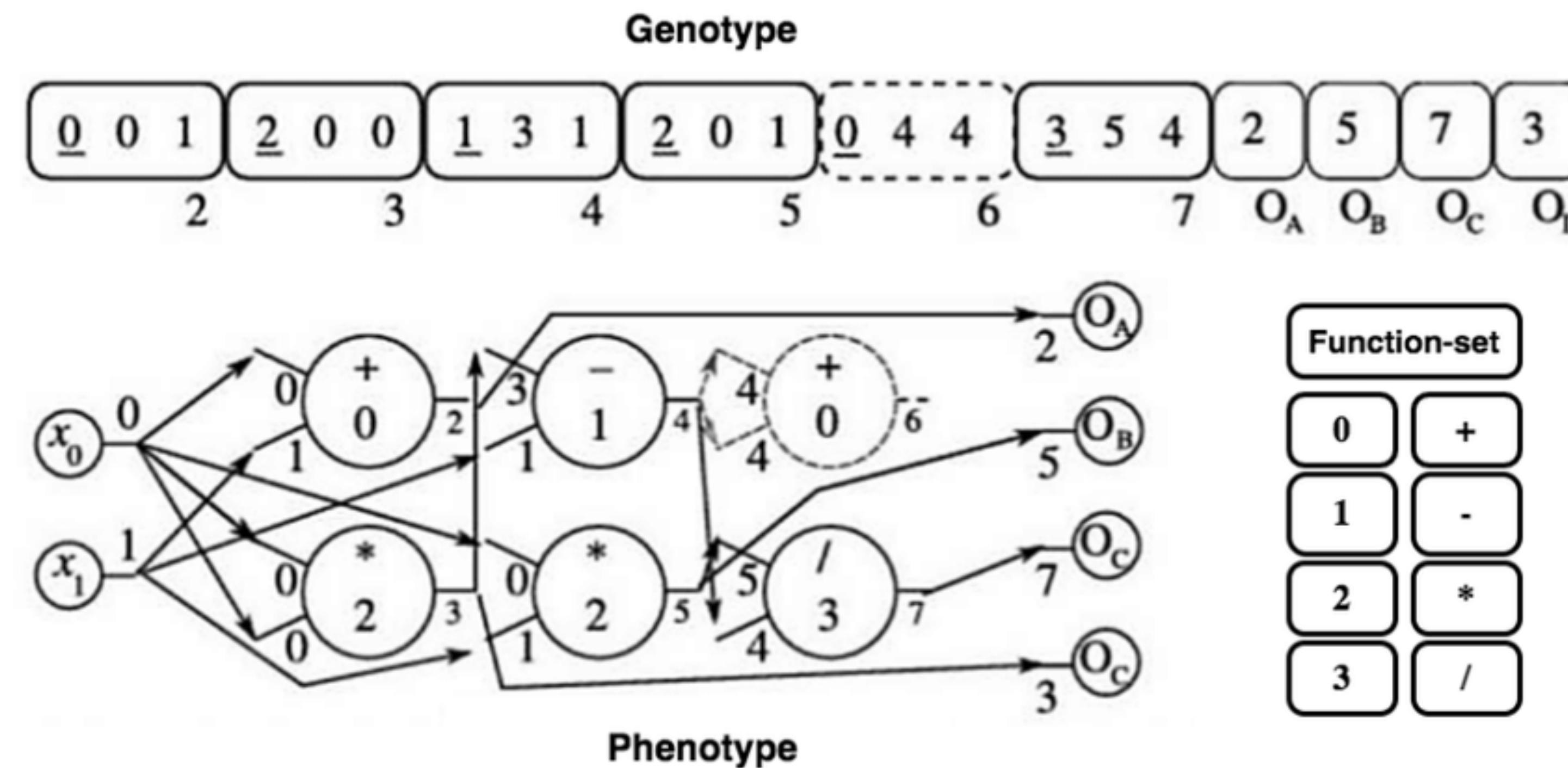


Figure: CGP Individual Encoding. Source: [6]

ShoreFor as a (sequential) CGP individual

ShoreFor
$\frac{dx}{dt} = c(F^+ + rF^-) + b$
$\Omega_{eq} = \frac{\sum_{i=1}^{2\phi} \Omega_i 10^{-i/\phi}}{\sum_{i=1}^{2\phi} 10^{-i/\phi}}$
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$\Omega = \frac{H_{s,b}}{wT_p}$

Table: ShoreFor system of equations

$$\Omega_{eq} = \frac{\sum_{i=1}^{2\phi} \Omega_i 10^{-i/\phi}}{\sum_{i=1}^{2\phi} 10^{-i/\phi}}$$

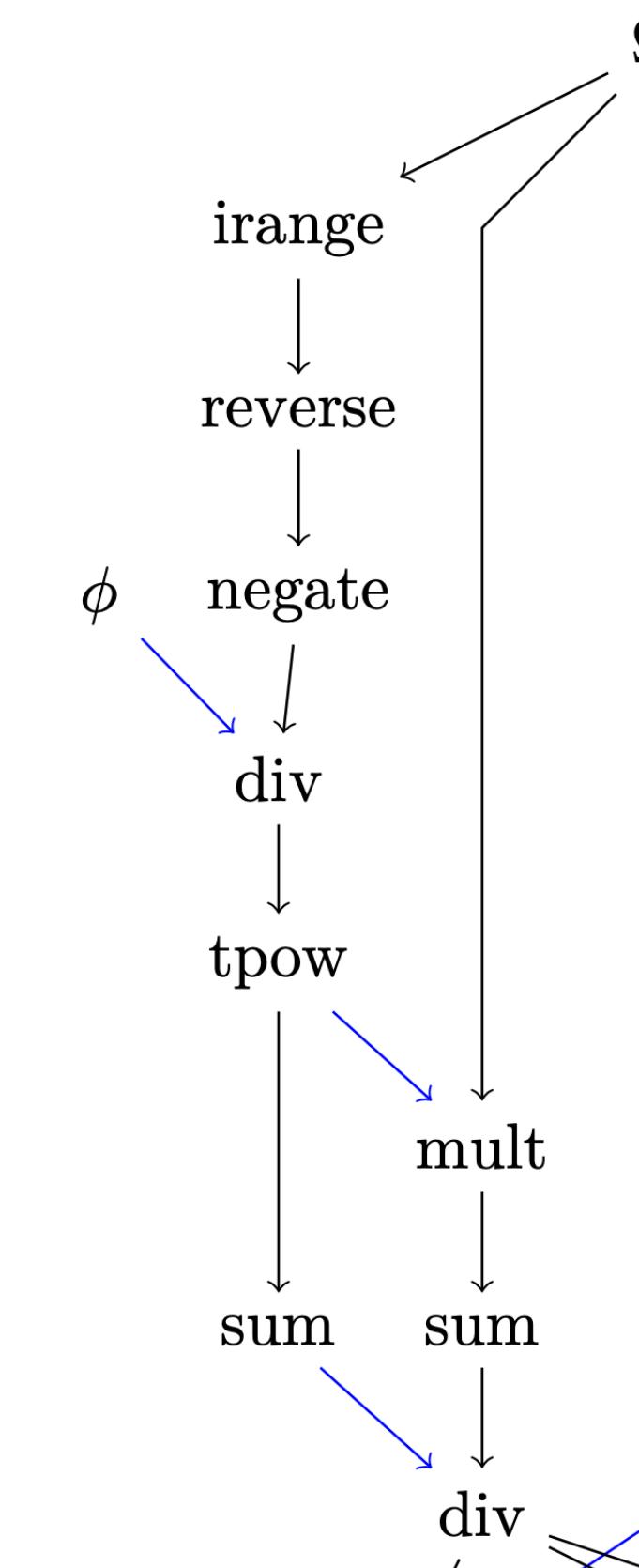
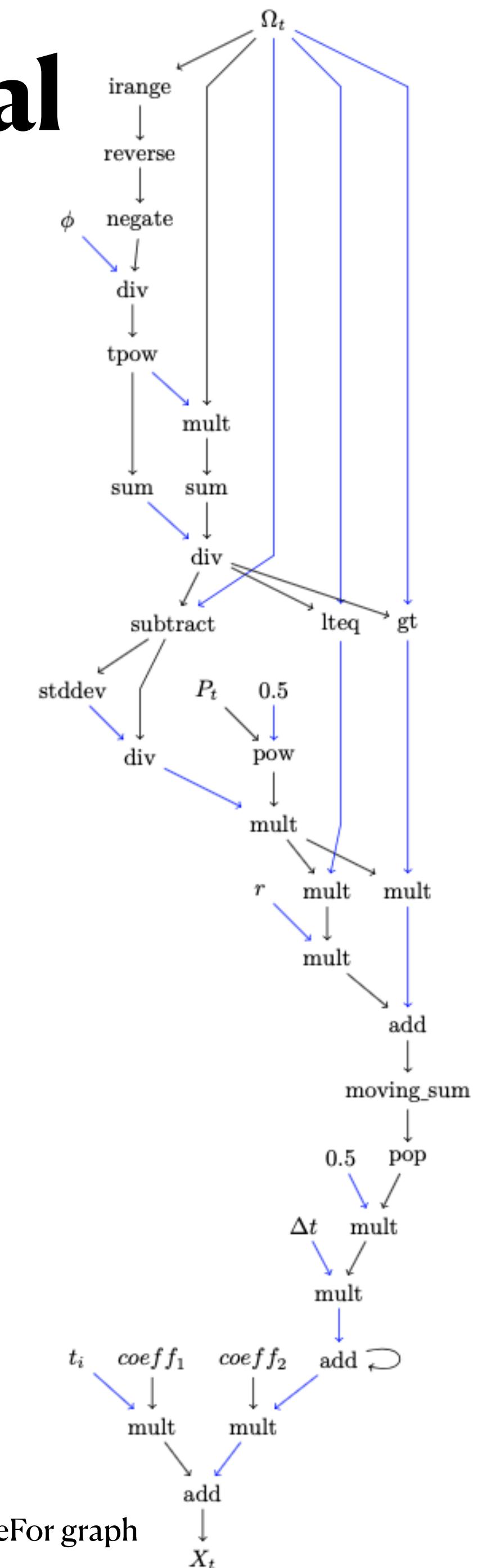


Figure: example sub-graph showing the encoding of the Ω_{eq} equation

Inputs
Ω
P
ϕ
0.5
dt_waves
current_time
coeff_time
c
r

Input window size = 2ϕ

Table, Figure: sequential ShoreFor graph



Experimental setup

Mielke fitness

$$\lambda = 1 - \frac{N^{-1} \sum_{i=1}^N (o_i - m_i)^2}{\sigma_o^2 + \sigma_m^2 + (\hat{o} - \hat{m})^2}$$

Function set	
Type	Description
List processing	list manipulation functions (e.g. tail, pop, reverse)
Scalar	Scalar operations (e.g. add, div, mult)
Vector	Vector operations (e.g. sum, mean, max)

Default configuration using $\mu+\lambda$ GA

Parameter	Value
Evaluation series length	11 years
Number of columns	300
Constraints	all
Population size	30/5
Mutation rate	0.1
Output mutation rate	0.3
Recurrent connection rate	0.1

Experiments

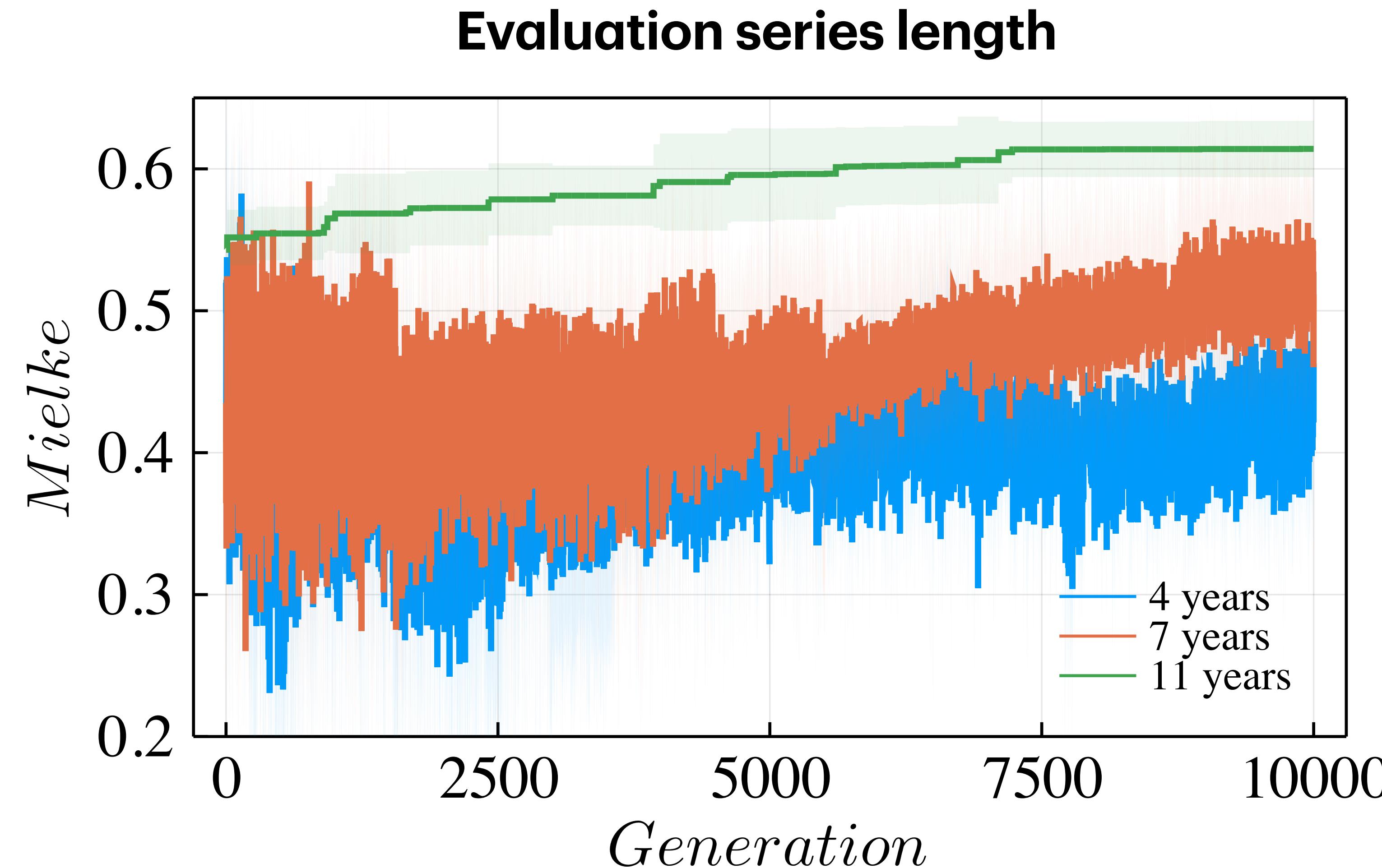


Figure: evaluation series length effect on model performance

Experiments

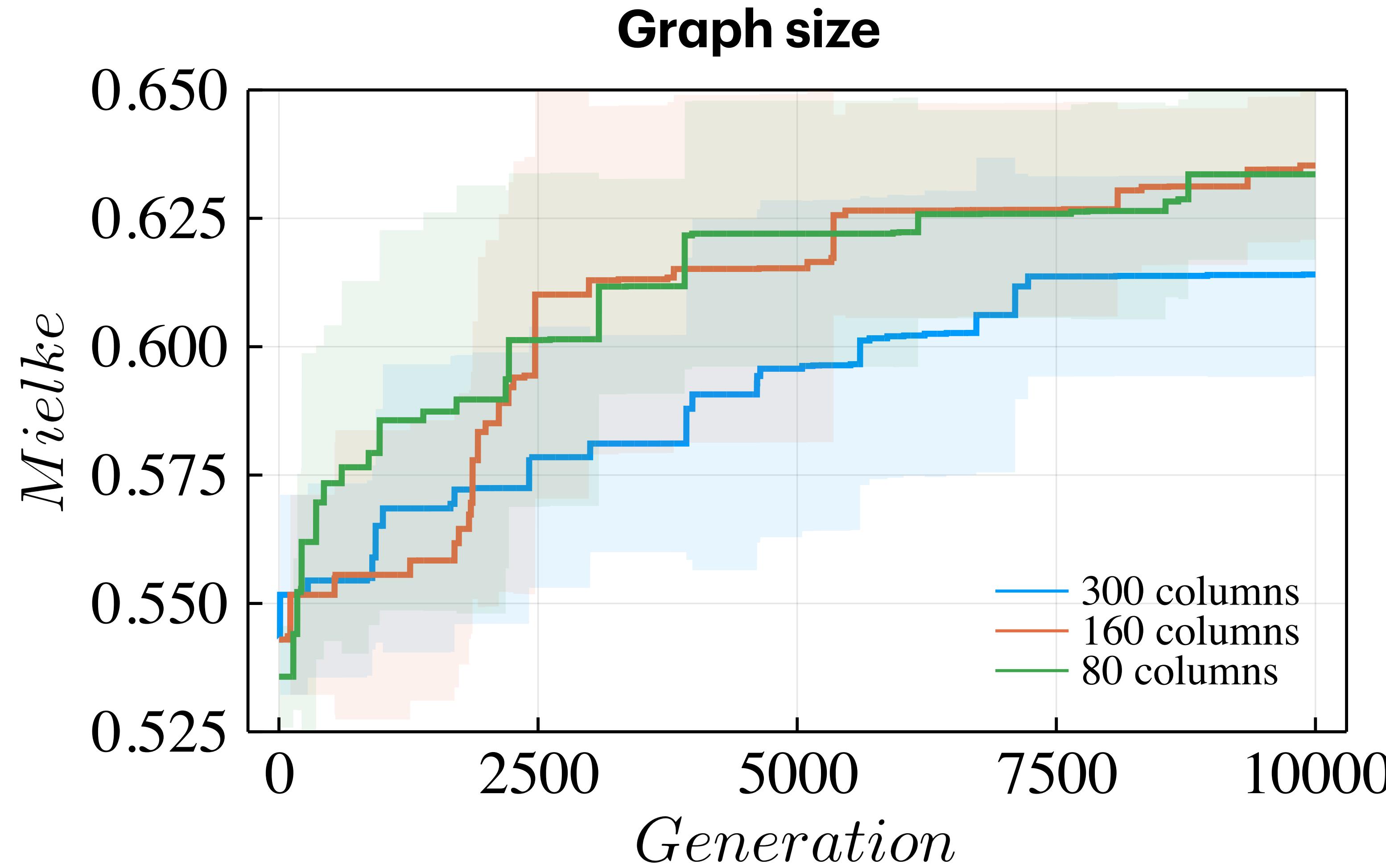


Figure: Sensitivity of CGP to different graph sizes

Experiments

Constraints

Level	Constraint
Evaluation	$std(\hat{y}) > \alpha_1$
Mutation	model length $> \alpha_2$
Mutation	$model(rand_1) \neq model(rand_2)$
Mutation	no input-output connections

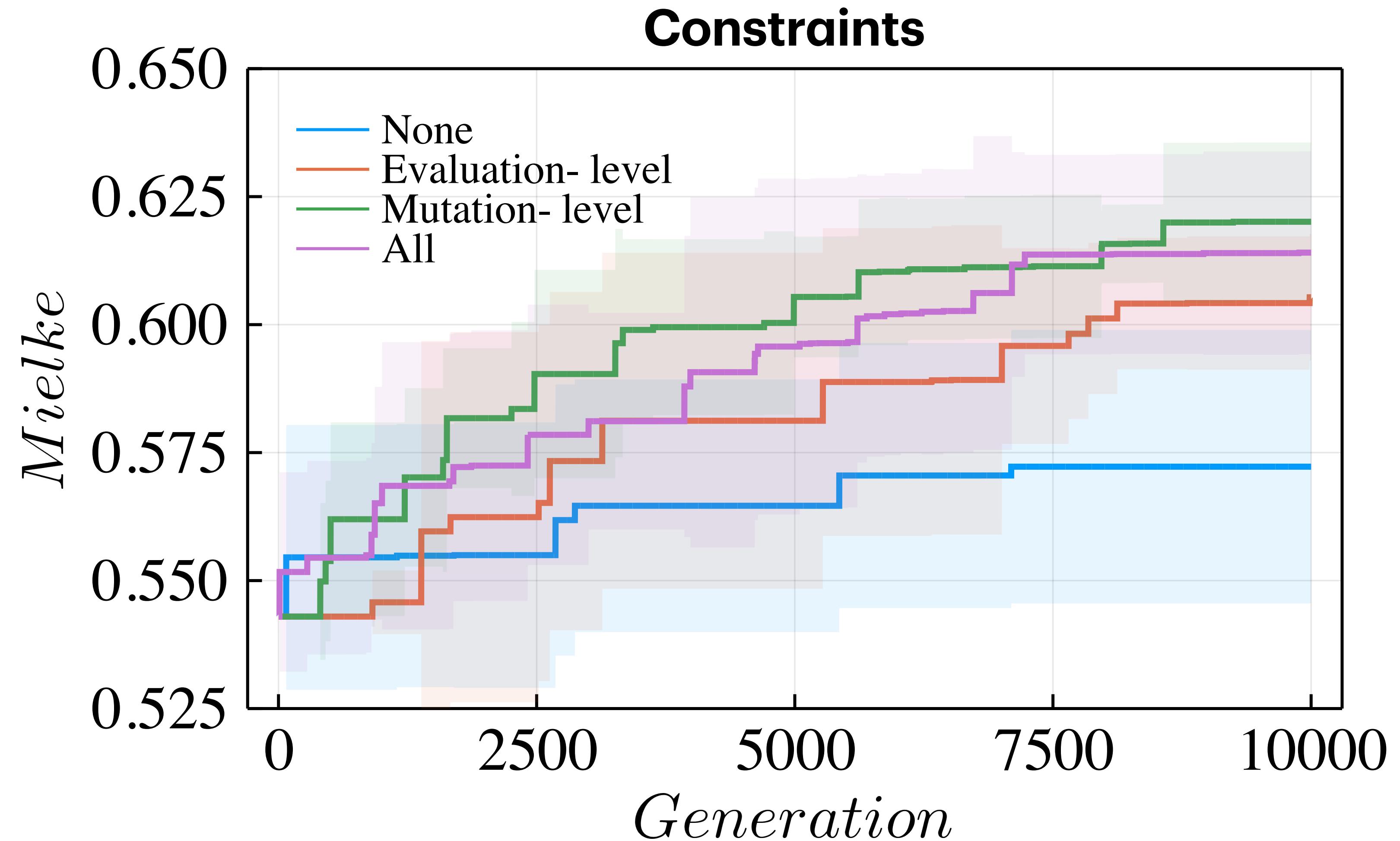


Figure: Effect of different constraint regimes on evolution

Experiments

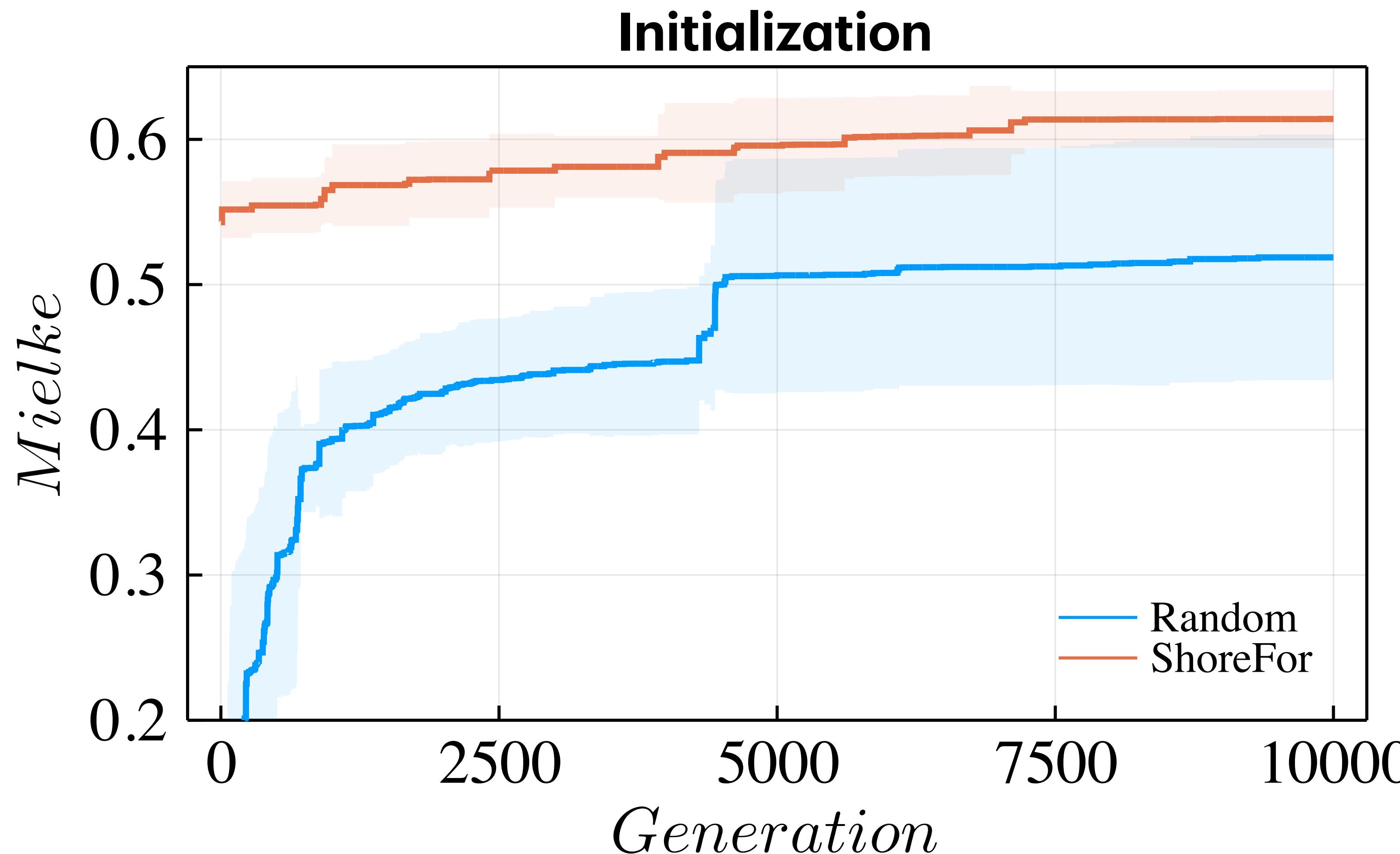
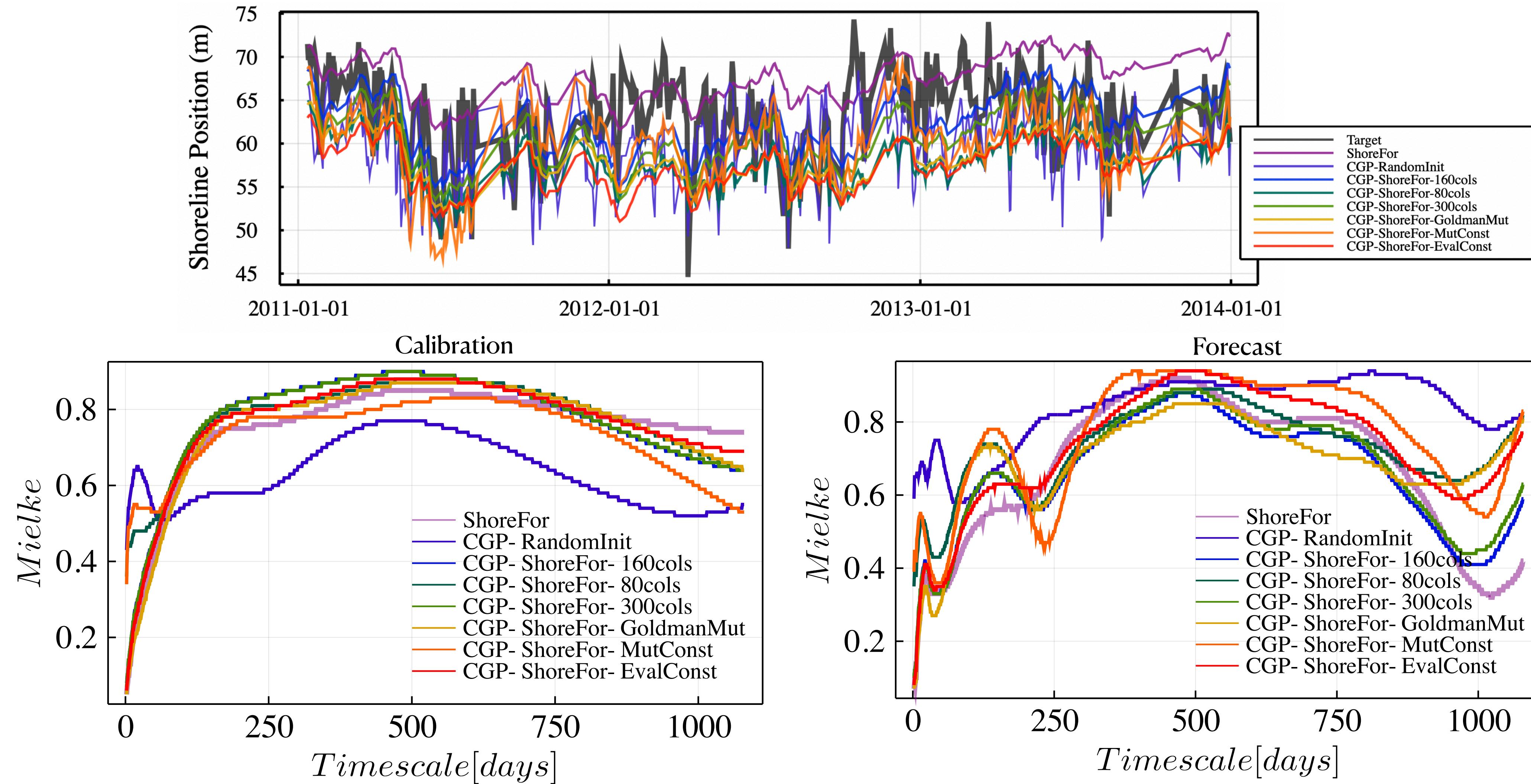


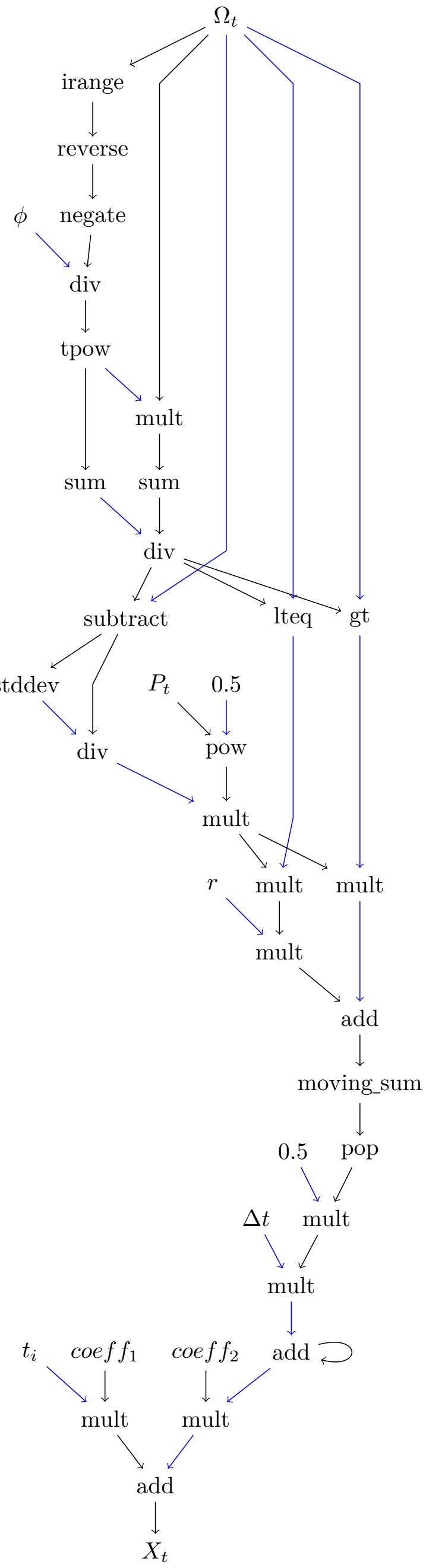
Figure: GI vs. pure CGP

Best individuals

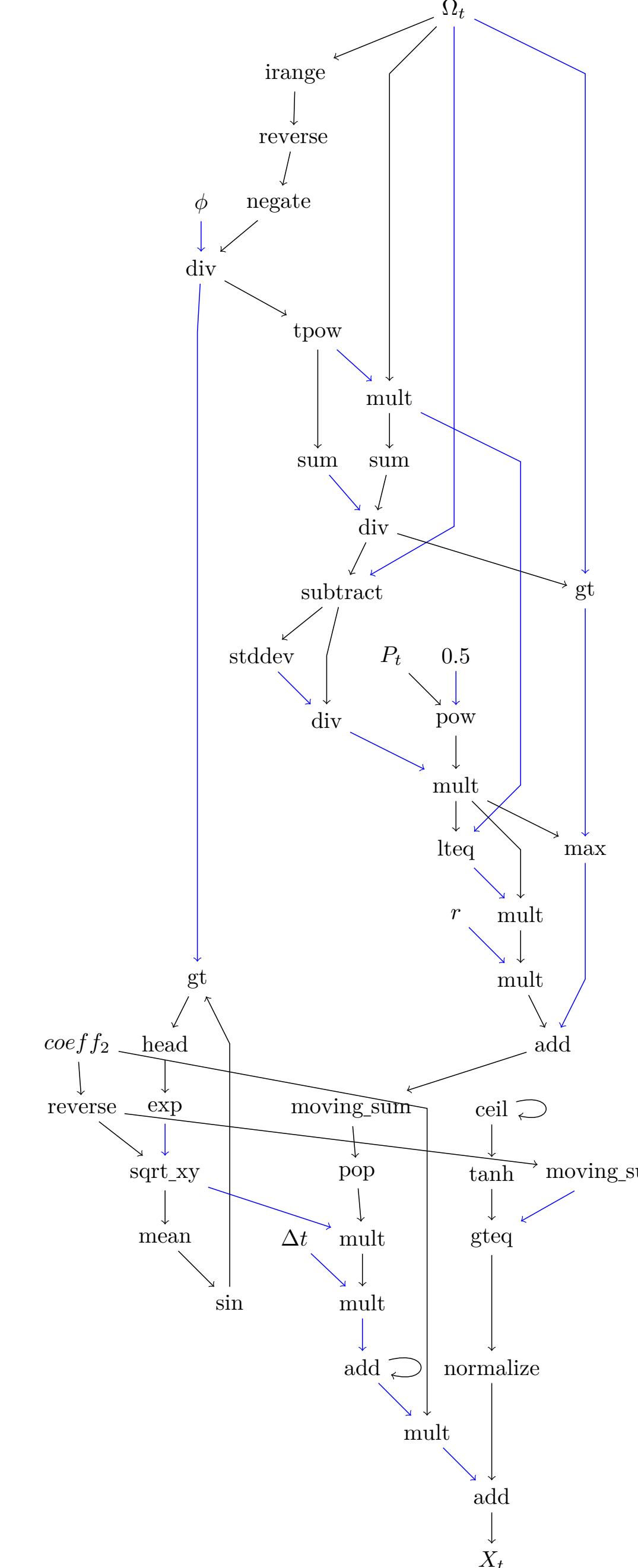
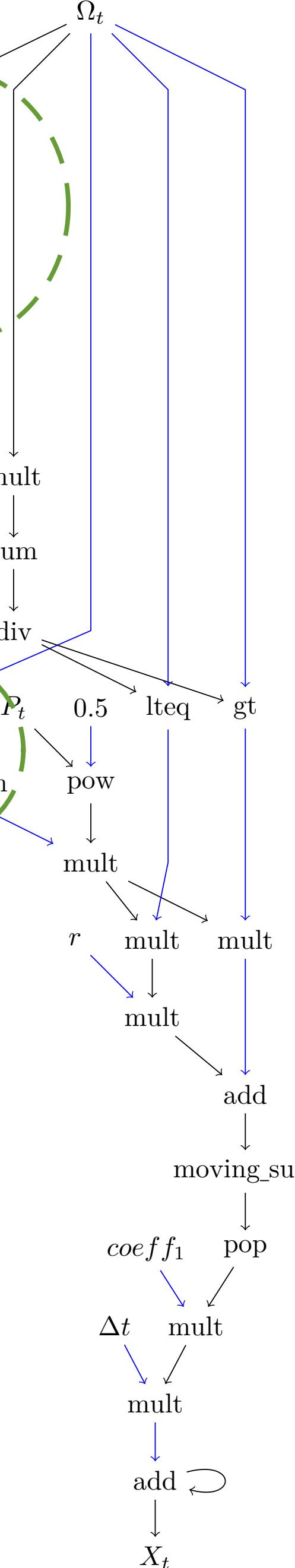


ShoreFor

CGP Inputs
Ω
P
ϕ
0.5
dt_waves
current_time
coeff1
coeff2
r

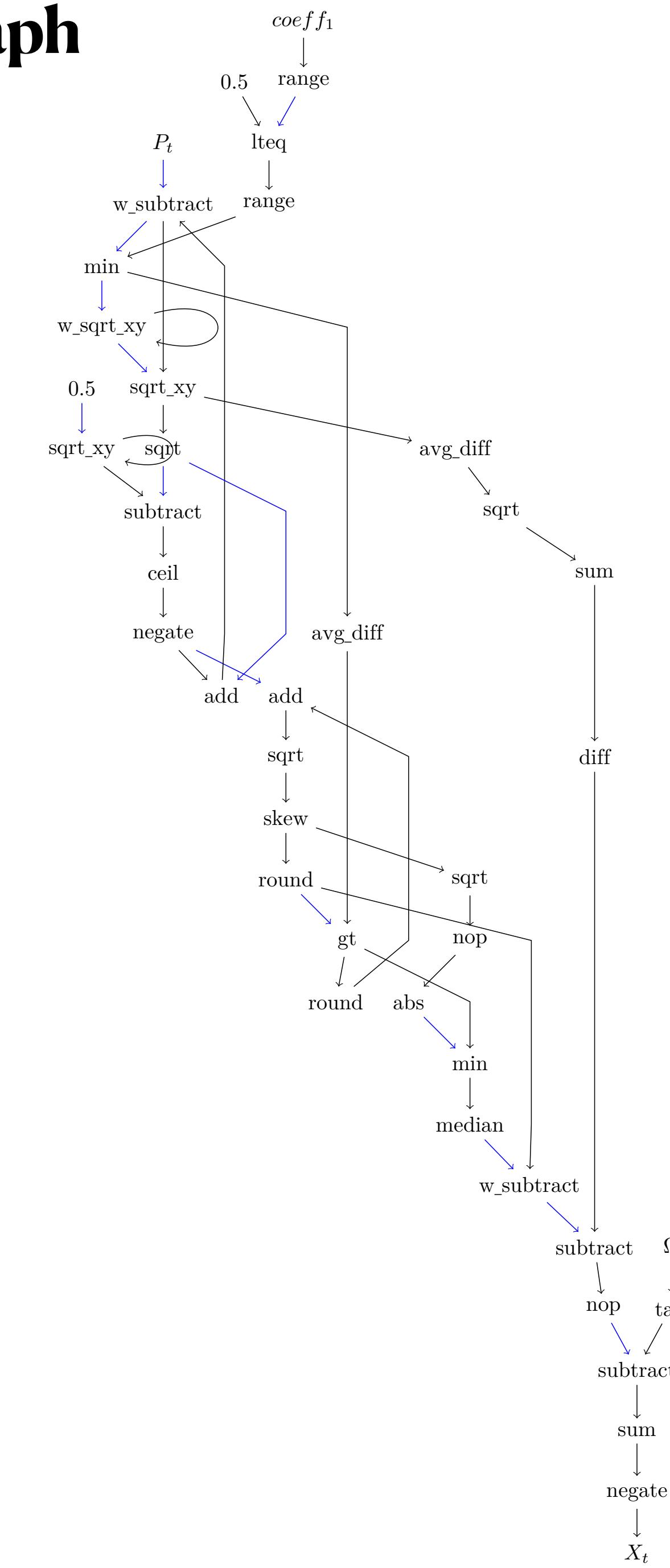
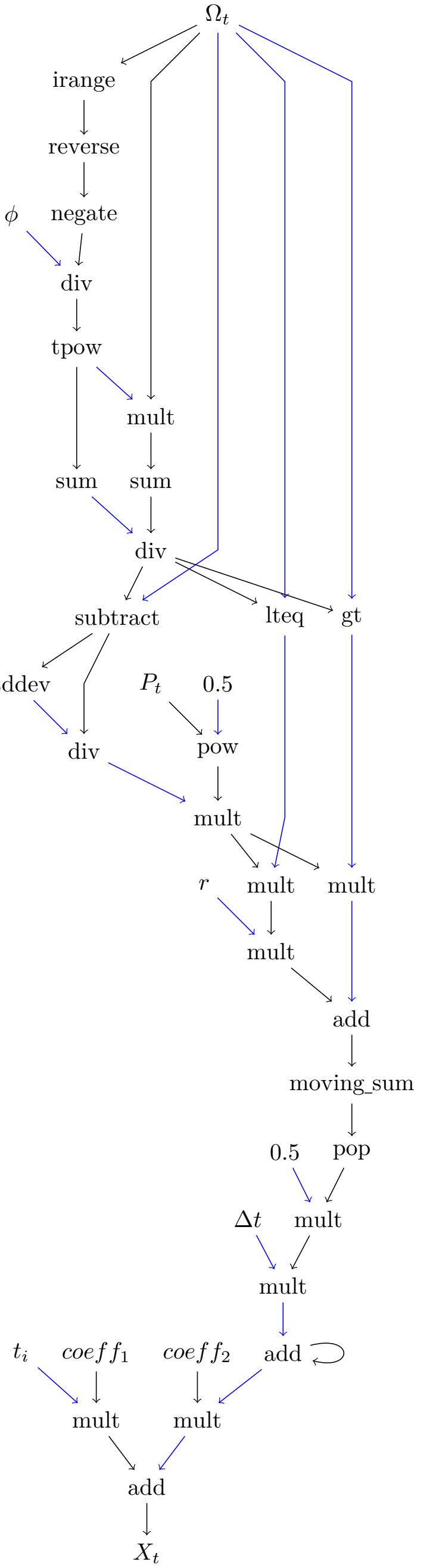


Evolved variants of ShoreFor



Evolved random-init graph

CGP Inputs
Ω
P
ϕ
0.5
dt_waves
current_time
coeff1
coeff2
r



Conclusions & future directions

Conclusions:

- First step towards the use of CGP in shoreline studies
- GI-based models demonstrated statistically higher skills
- Promising first results, establishing basis for future work

Future perspectives:

- Sequential (RCCGP) vs full-series formulations (CGP)
- Additional sites/datasets for multi-objective optimization

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Thank you

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