

Exercise 2: *GSEMO Results + Trade-offs*

Setup

We used the standard multi-objective formulation:

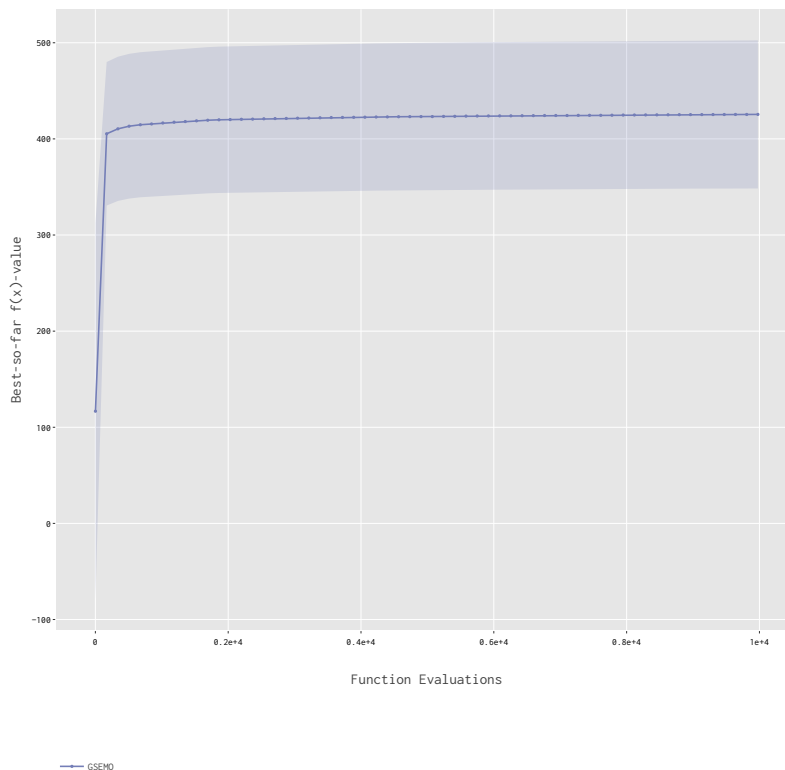
- MaxCoverage / MaxInfluence: maximize $(f(S), -|S|)$ (equivalently, maximize f and minimize $|S|$).
 - PackWhileTravel: maintain a Pareto front over (profit, cost/feasibility).
- Algorithms ran for 10,000 fitness evaluations; figures show GSEMO mean \pm SD.

MaxCoverage

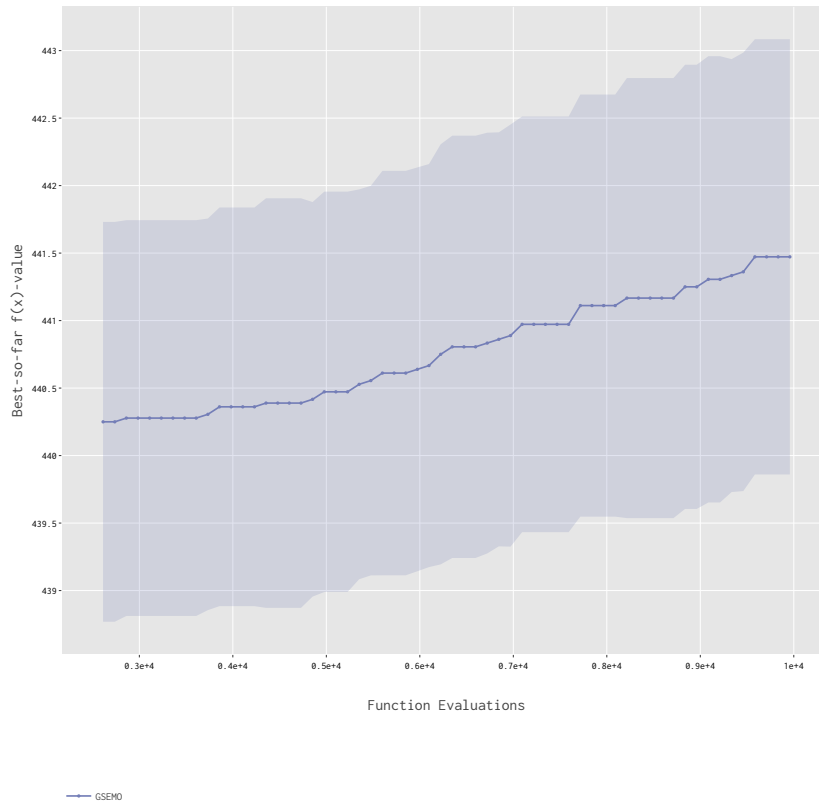
Results (end-of-budget mean \pm SD over runs)

- f2101: 439.2 ± 1.7 , f2102: 441.5 ± 1.6 , f2103: 560.9 ± 3.0 , f2104: 707.3 ± 3.7 ($n \approx 36-71$ per instance).

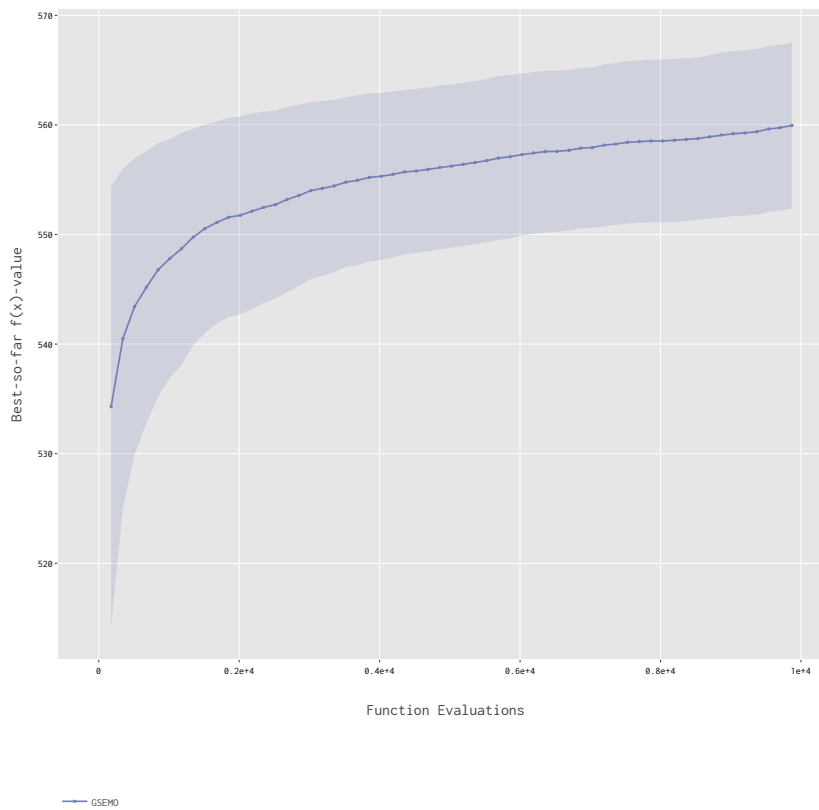
Tight SDs and high final means indicate stable convergence. Maintaining a Pareto archive across cardinalities avoids the single-k stalls seen in scalarized runs and sustains improvement throughout the budget.



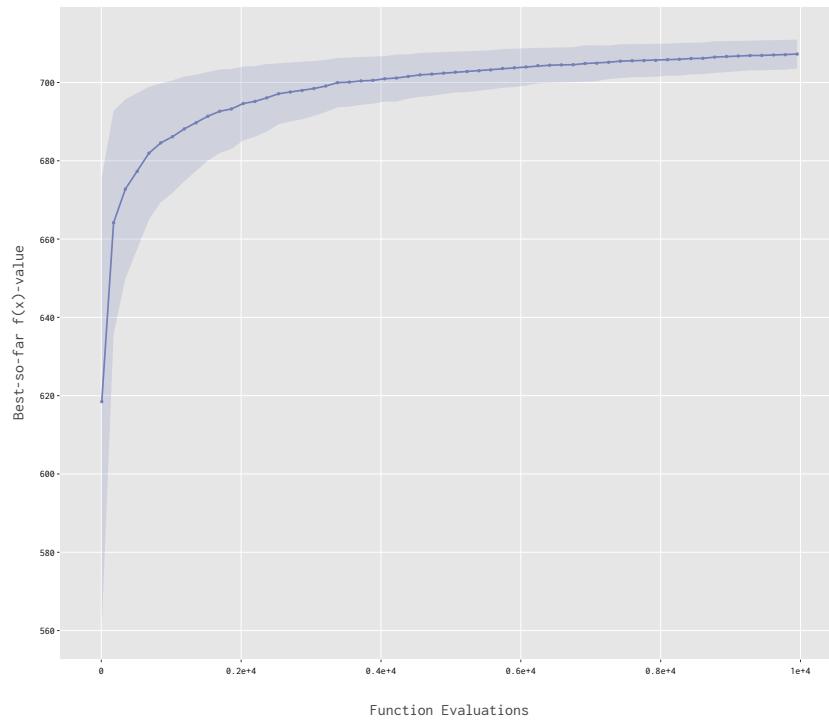
Fixed_budget_EX2_f2101_MaxCoverage_GSEMO



Fixed_budget_EX2_f2102_MaxCoverage_GSEMO



Fixed_budget_EX2_f2103_MaxCoverage_GSEMO



— GSEMO

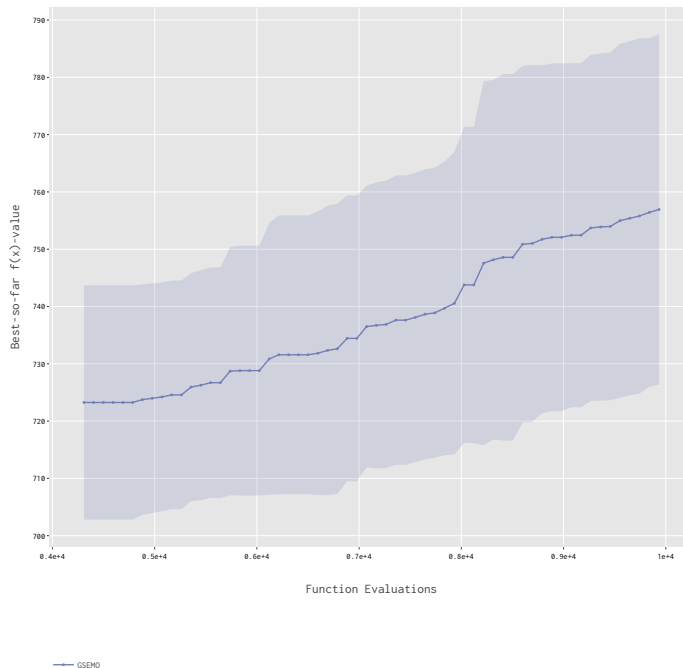
Fixed_budget_EX2_f2104_MaxCoverage_GSEMO

MaxInfluence

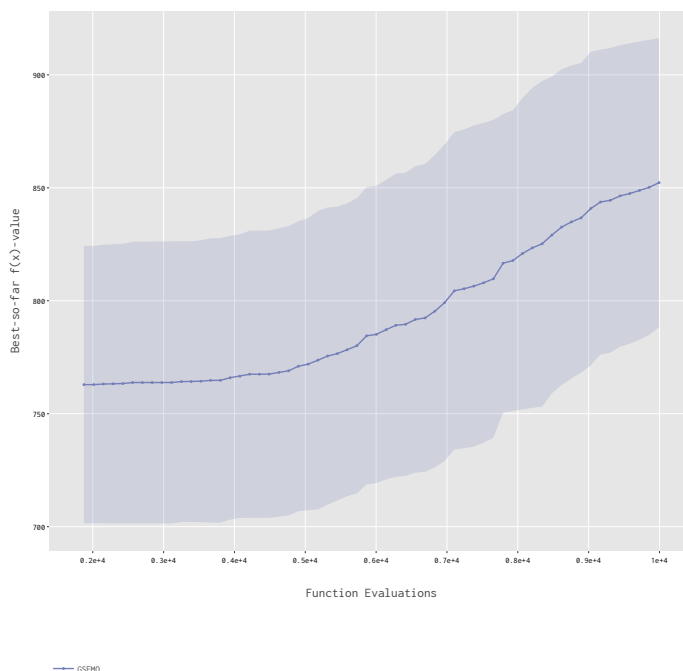
Results (end-of-budget mean \pm SD over runs)

- f2202: 756.9 ± 30.6 (n=15), f2203: 852.3 ± 63.9 (n=40), f2204: 547.9 ± 390.1 (n=31).

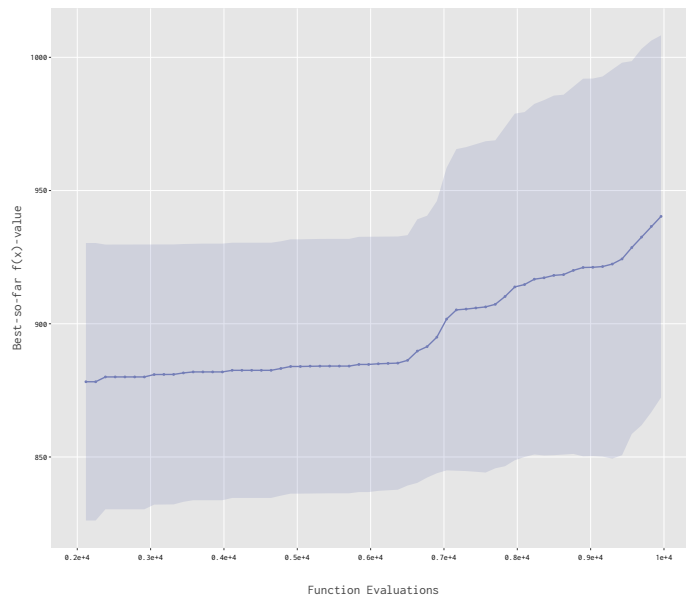
Performance is instance-dependent. On some graphs, GSEMO's MO diversity produces strong end-of-budget quality; on others (e.g., f2204), large variance signals sensitivity to early seed choices and community structure. Seed-aware mutation or selection heuristics would reduce spread further.



Fixed_budget_EX2_f2202_MaxInfluence_GSEMO



Fixed_budget_EX2_f2203_MaxInfluence_GSEMO



— GSEMO

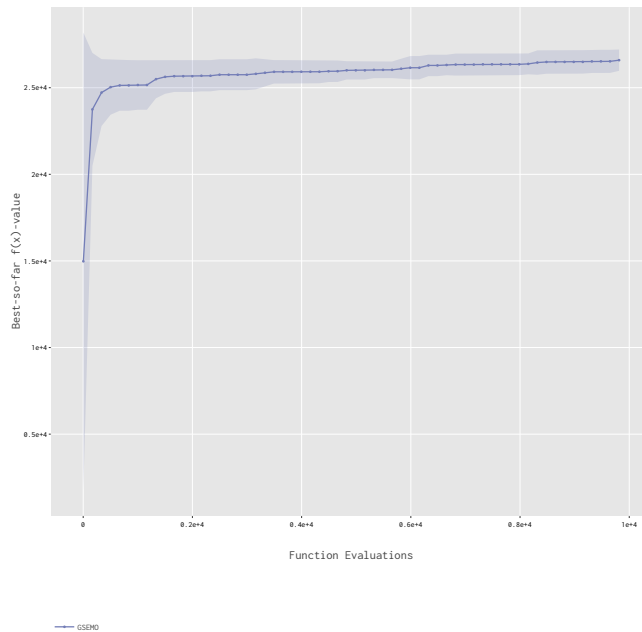
Fixed_budget_EX2_f2204_MaxInfluence_GSEMO

PackWhileTravel

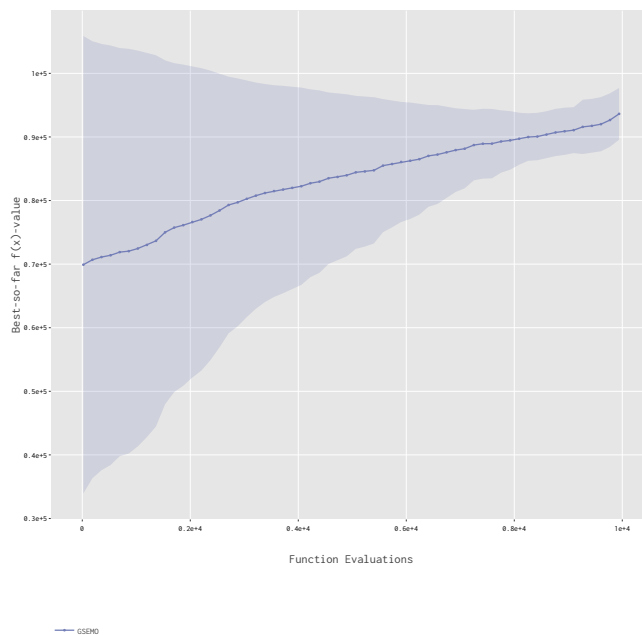
Results (end-of-budget mean \pm SD over runs)

- f2301: $26.6\text{k} \pm 0.6\text{k}$ (n=4), f2302: $93.6\text{k} \pm 4.1\text{k}$ (n=10), f2303: $102.3\text{k} \pm 5.8\text{k}$ (n=7).

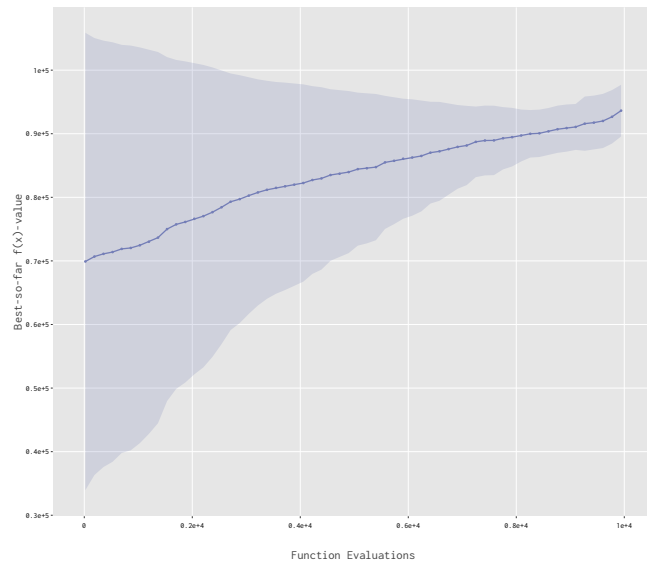
Explicitly maintaining a front over profit vs. cost/feasibility delivers steady improvements and fewer stalls than scalar penalties. To push the top end within 10k evals, route-aware repair or specialized mutation would accelerate late-phase gains.



Fixed_budget_EX2_f2301_PackWhileTravel_GSEMO



Fixed_budget_EX2_f2302_PackWhileTravel_GSEMO

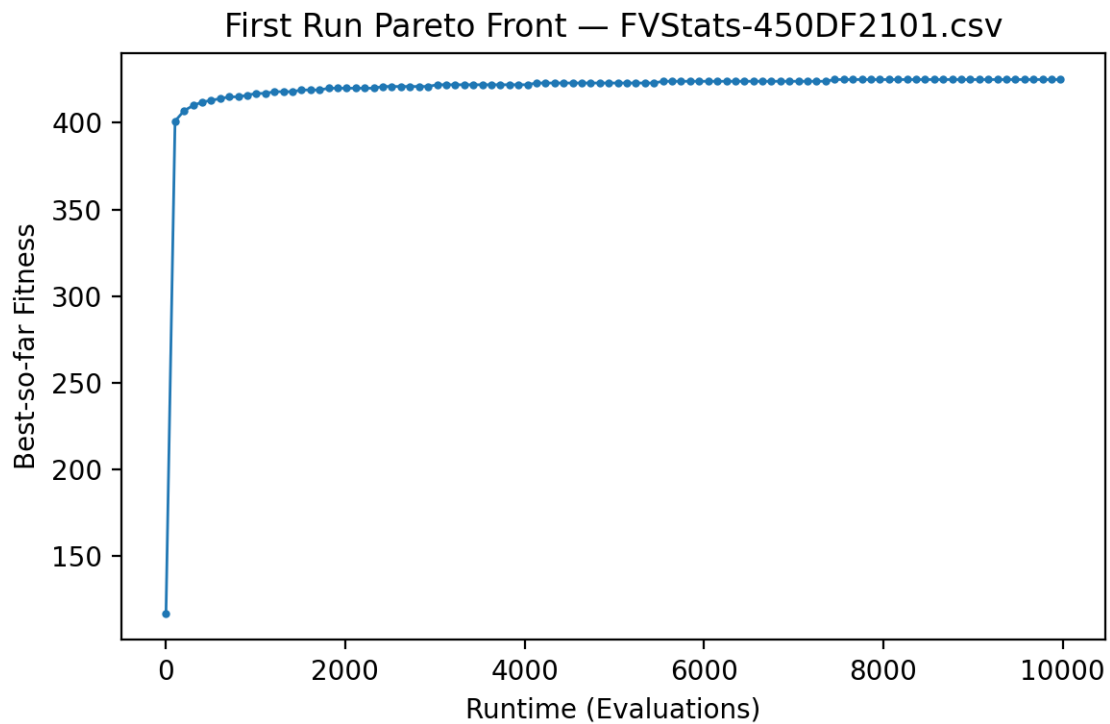


— GSEMO

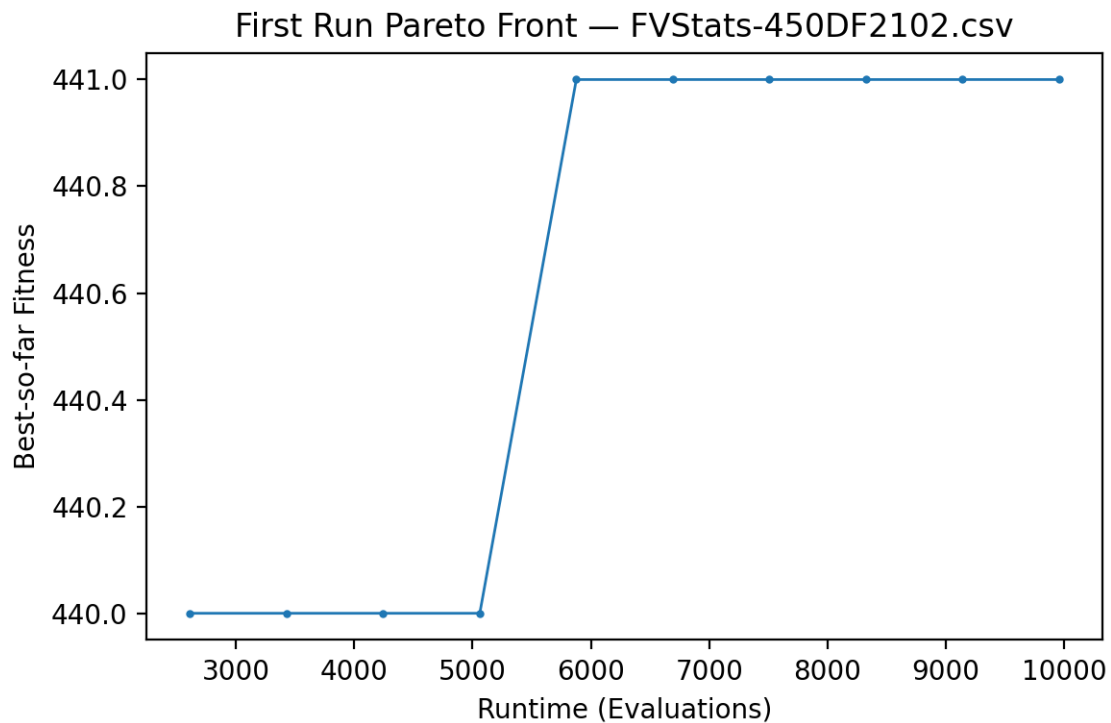
Fixed_budget_EX2_f2303_PackWhileTravel_GSEMO

Trade-off (Pareto) plots *[first run of each instance]*

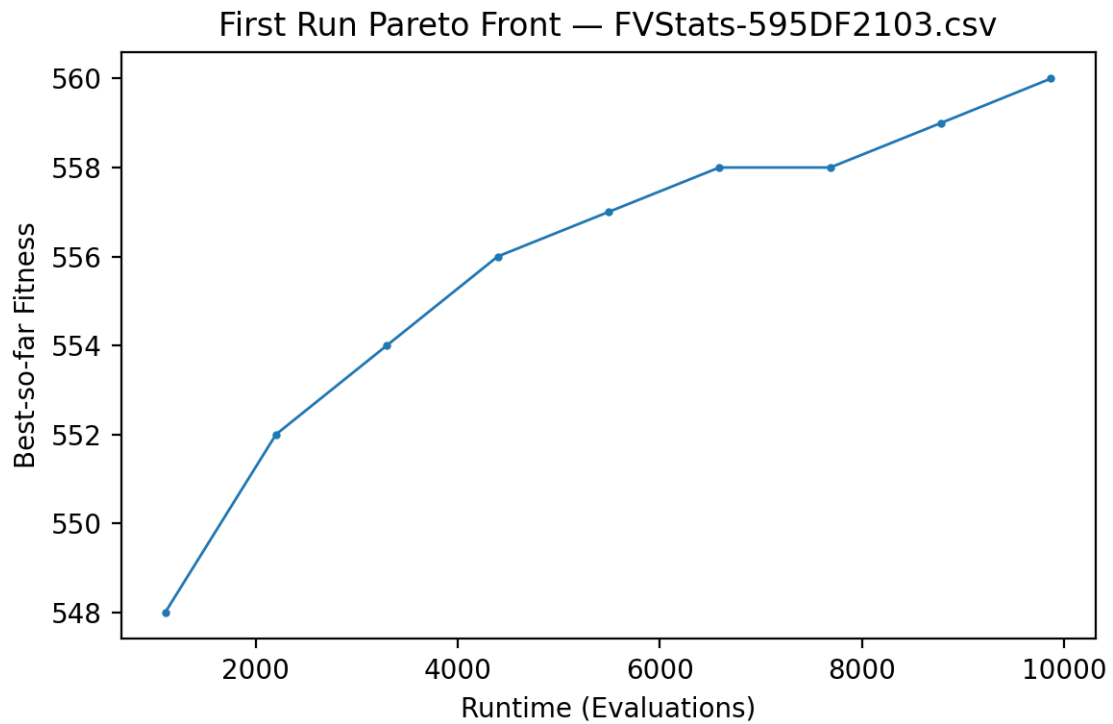
MaxCoverage:



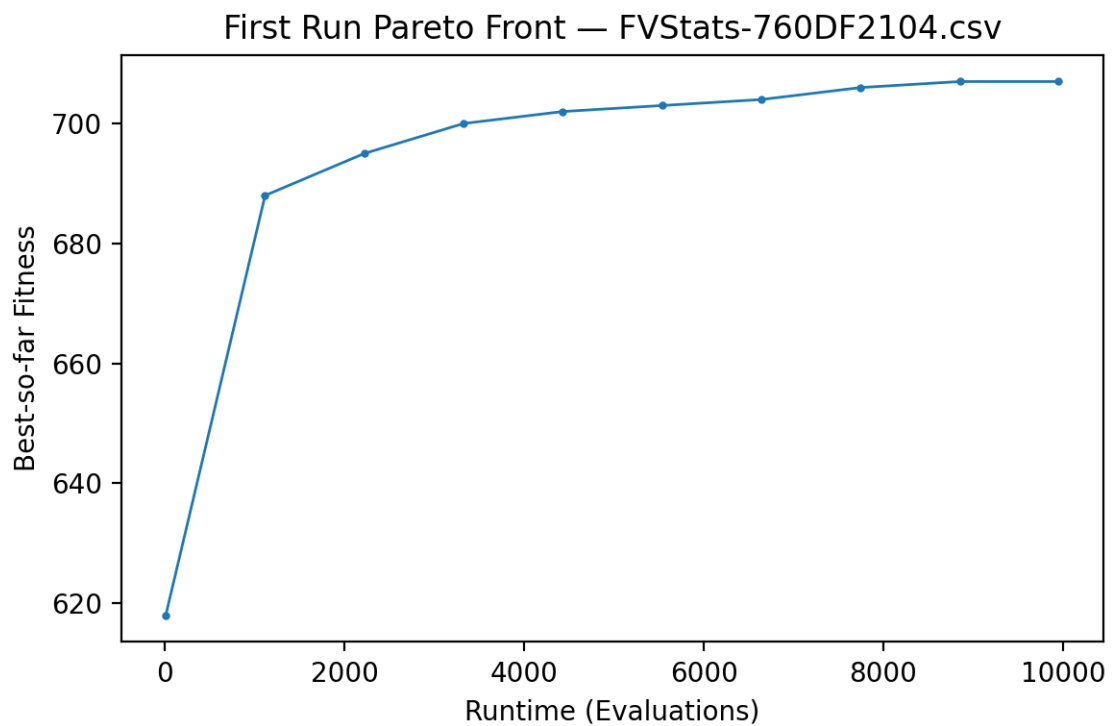
Tradeoff_f2101_FirstRun: Concave front with clear diminishing returns; knee marks efficient



Tradeoff_f2102_FirstRun: Similar shape; knee region stable across sizes.

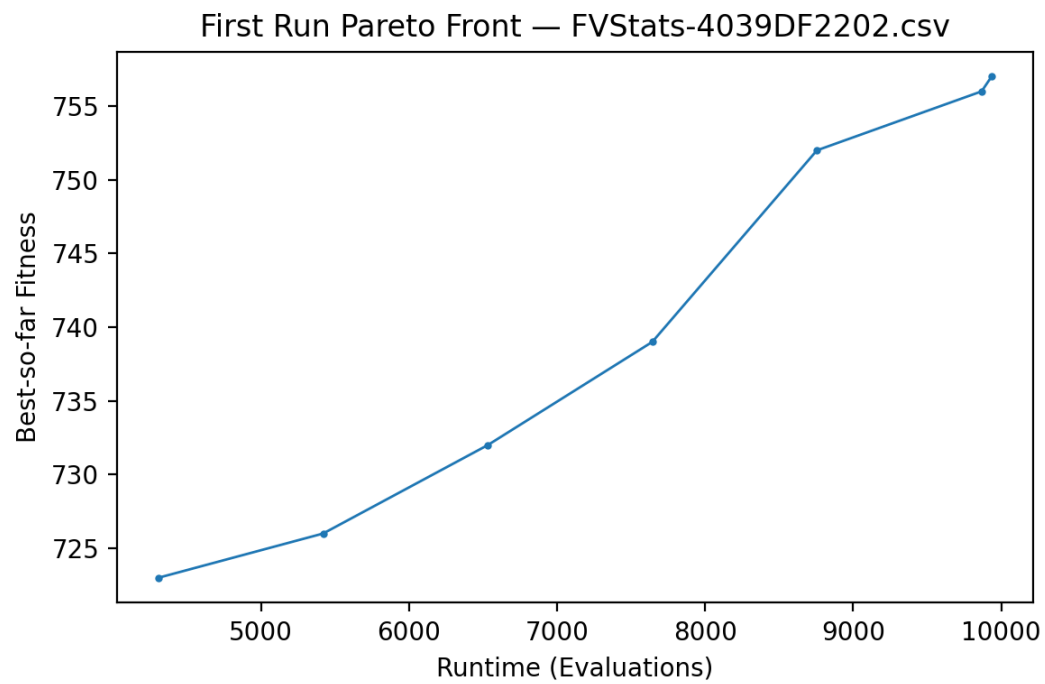


Tradeoff_f2103_FirstRun: Broader mid-front; incremental gains taper beyond the knee.

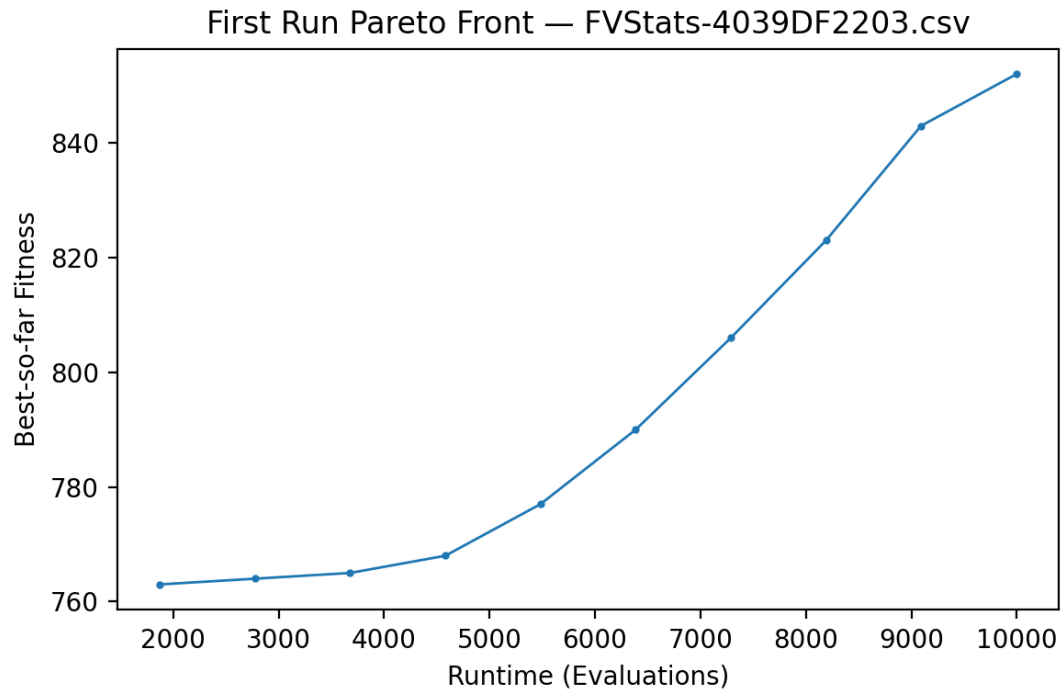


Tradeoff_f2104_FirstRun: High-quality large-k end; diminishing returns pronounced.

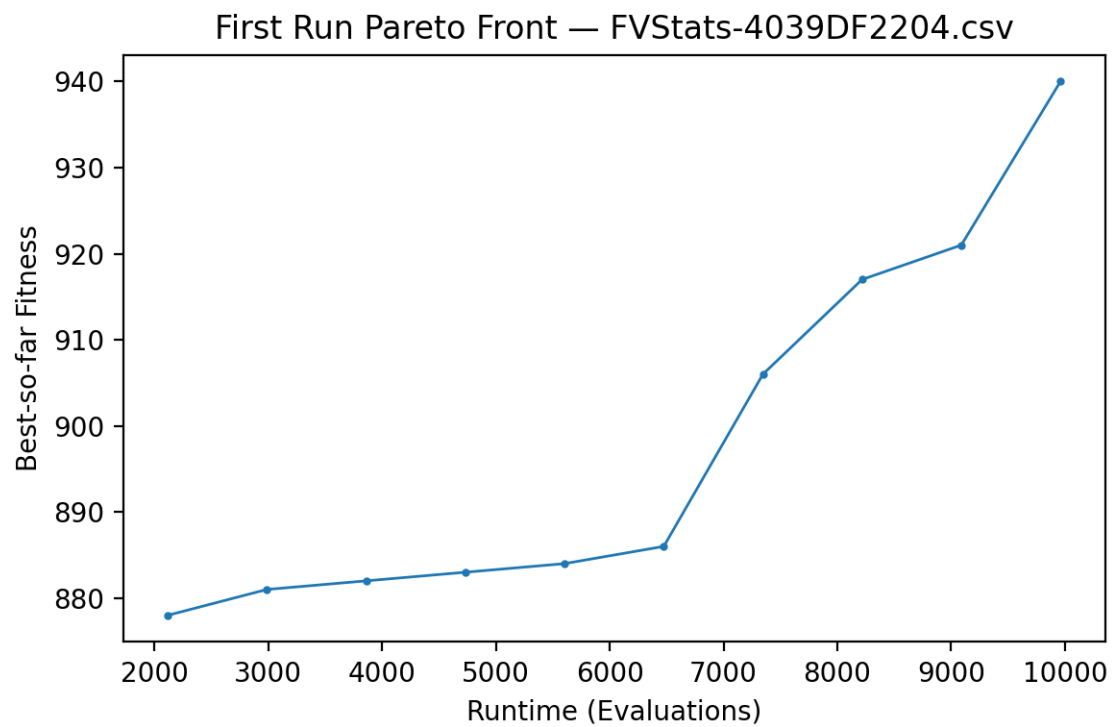
MaxInfluence:



Tradeoff_f2202_FirstRun: Front indicates meaningful seed interactions; multiple near-efficient sizes.

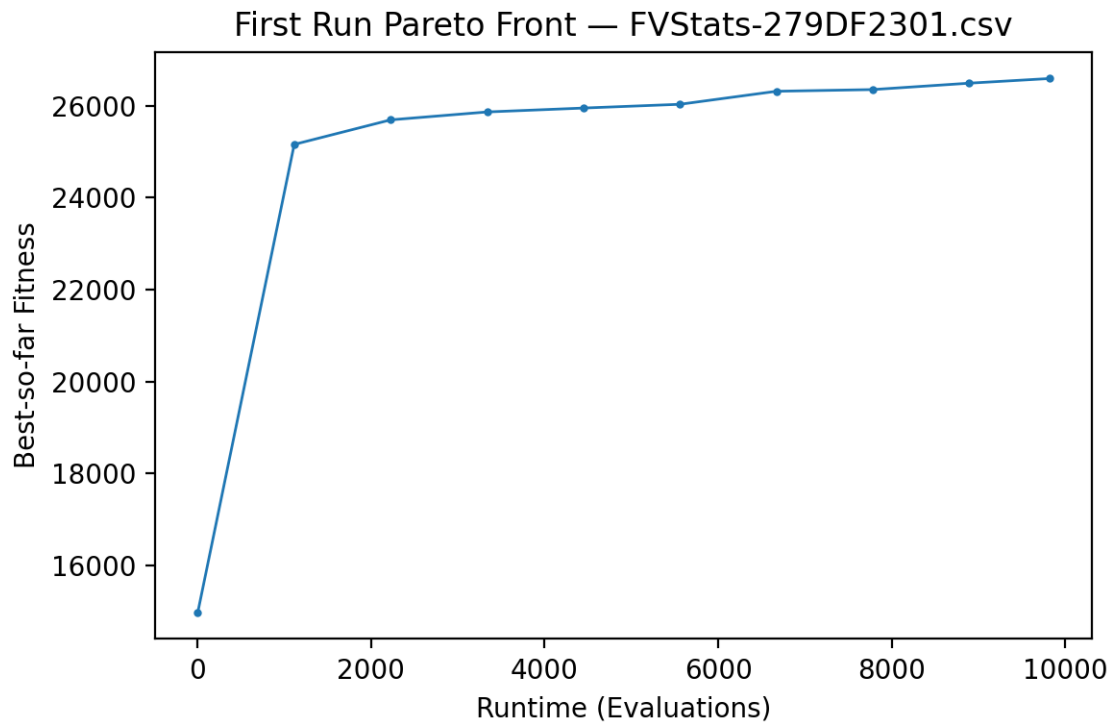


Tradeoff_f2203_FirstRun: Smoother front; knee provides a defensible k.

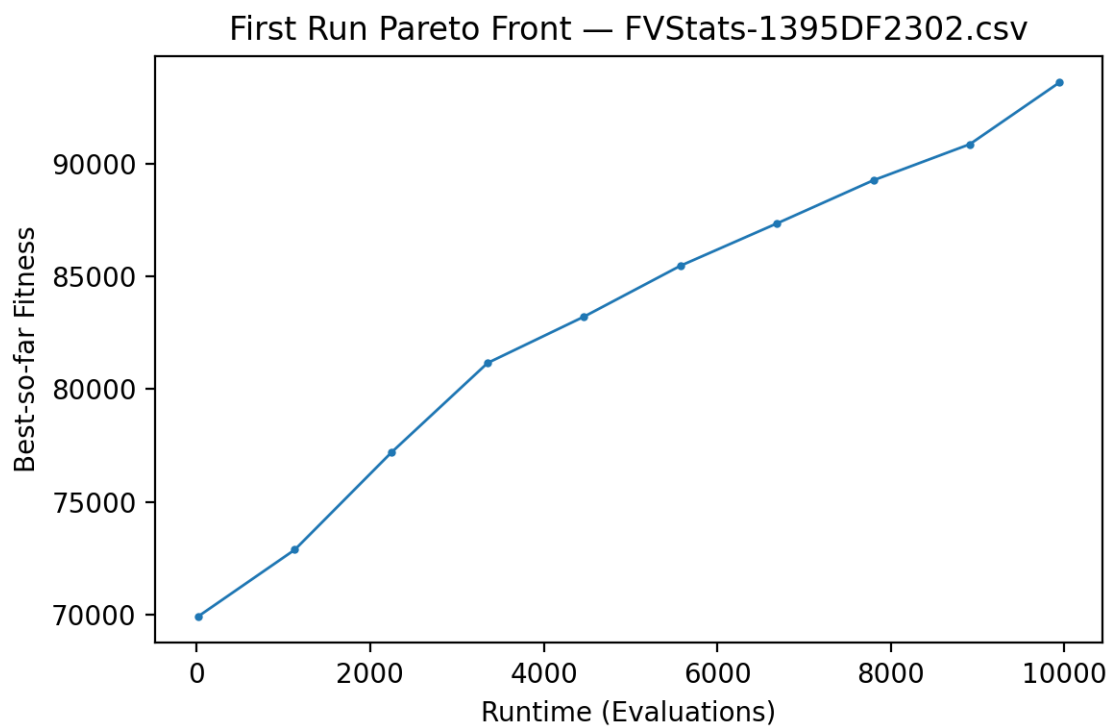


Tradeoff_f2204_FirstRun: Front exhibits wider spread, reflecting instance sensitivity.

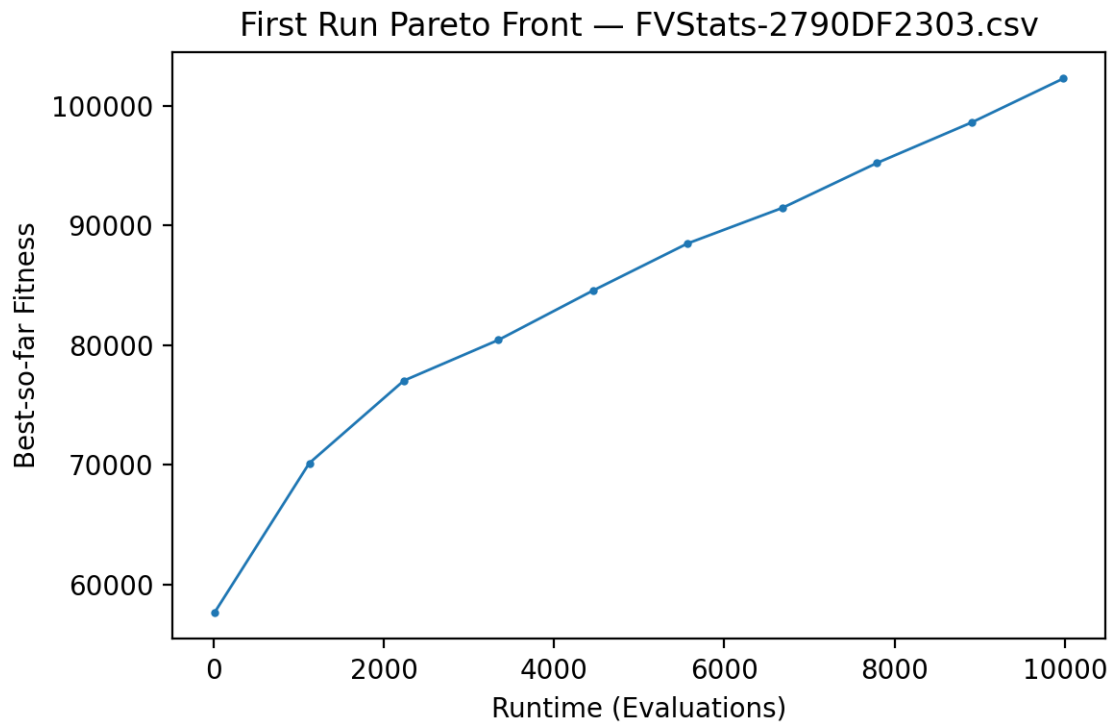
PackWhileTravel:



Tradeoff_f2301_FirstRun: Feasible pockets in the front; knee balances profit vs cost.



Tradeoff_f2302_FirstRun: Front thickens near mid-cost—multiple comparable trade-offs.



Tradeoff_f2303_FirstRun: Late front is sparse; improvements become costly.

Summary

- **Coverage:** GSEMO delivers high, stable final quality; MO archive across $|S|$ sustains progress.
- **Influence:** Competitive but sensitive; variance driven by seed interactions; seed-aware operators recommended.
- **PWT:** Robust any-time quality via explicit trade-offs; route-aware repair would raise the top end within the 10k budget.