

# HNCO

## Influence of the learning rate on the performance of PBIL

November 4, 2017

### Abstract

PBIL is applied many times to the same collection of fitness functions (bit vector size  $n = 100$ ), each time with a different learning rate taken from a finite set of values. All learning rates are ranked according to their median fitness over 20 independent runs, first for each fitness function, then across the entire collection of fitness functions. The mean and standard deviation of fitness are also plotted as a function of the learning rate.

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# 1 Plan

```
{
  "exec": "hnco",
  "opt": "-s 100 --map 1 --map-random -i 0 -b 200000 --print-performance",
  "num_runs": 20,
  "parallel": false,
  "results": "results",
  "graphics": "graphics",
  "report": "report",
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    "id": "learning-rate",
    "values": [ 1e-2, 2e-2, 5e-2, 1e-1, 2e-1, 5e-1, 1 ],
    "logscale": true,
    "boxwidth": "$1 * 0.3"
  },
  "functions": [
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      "id": "one-max",
      "opt": "-F 0 --stop-on-maximum",
      "col": ">{\\nprounddigits{0}}N{3}{0}"
    },
    {
      "id": "leading-ones",
      "opt": "-F 10 --stop-on-maximum",
      "col": ">{\\nprounddigits{0}}N{3}{0}"
    },
    {
      "id": "jmp-5",
      "opt": "-F 30 --stop-on-maximum -t 5",
      "col": ">{\\nprounddigits{0}}N{3}{0}"
    },
    {
      "id": "nk",
      "opt": "-F 60 -p instances/nk.100.4",
      "col": ">{\\nprounddigits{2}}N{1}{2}"
    },
    {
      "id": "max-sat",
      "opt": "-F 70 -p instances/ms.100.3.1000 --cache",
      "col": ">{\\nprounddigits{0}}N{3}{0}"
    },
    {
      "id": "labs",
      "opt": "-F 80",
      "col": ">{\\nprounddigits{2}}N{1}{2}"
    },
    {
      "id": "ep",
      "opt": "-F 90 -p instances/ep.100",
      "reverse": true,
      "logscale": true,
      "col": ">{\\nprounddigits{1}}N{1}{1}"
    },
    {
      "id": "cancel",
      "opt": "-F 100 -s 99",
      "reverse": true,
      "col": ">{\\nprounddigits{2}}N{1}{2}"
    },
    {
      "id": "walsh2",
      "opt": "-F 162 -p instances/walsh2.100 --cache",

```

```

        "col": ">{\nprounddigits{2}}N{3}{2}"
    }
],
"algorithms": [
    {
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        "opt": "-A 500 -x 10 -y 1"
    }
]
}

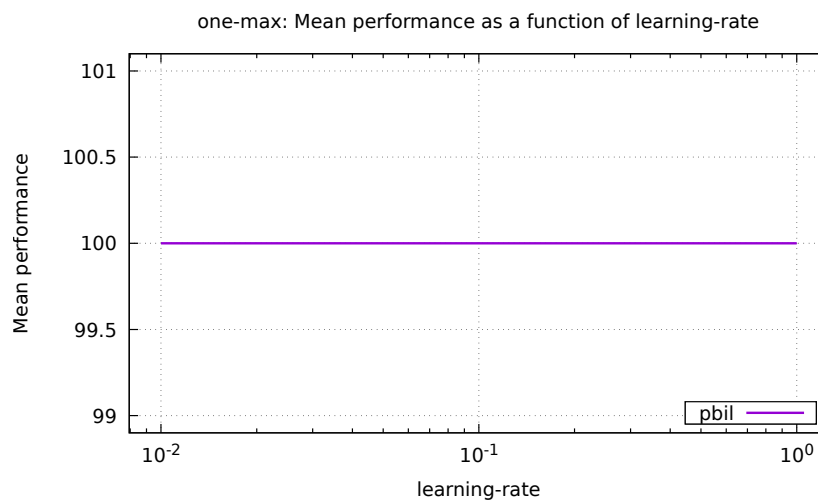
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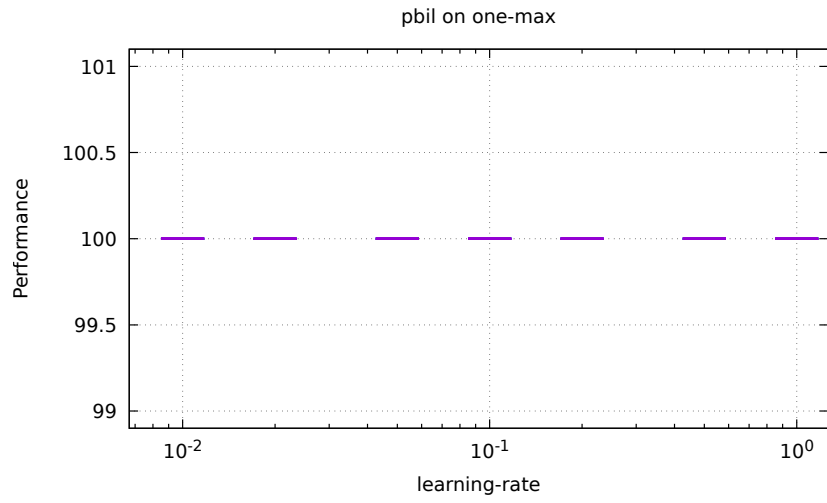
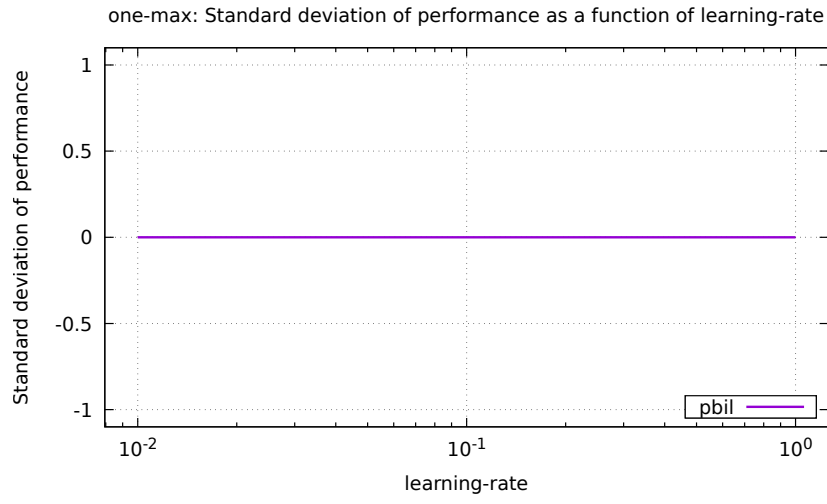
## 2 Rankings

algorithm	learning-rate	rank distribution						
		1	2	3	4	5	6	7
pbil	1	7	0	1	0	0	1	0
pbil	0.01	4	1	2	1	1	0	0
pbil	0.02	3	3	1	1	0	0	1
pbil	0.05	3	0	0	2	0	3	1
pbil	0.5	2	2	0	1	2	2	0
pbil	0.2	2	0	1	1	4	1	0
pbil	0.1	2	0	1	1	0	1	4

## 3 Function one-max

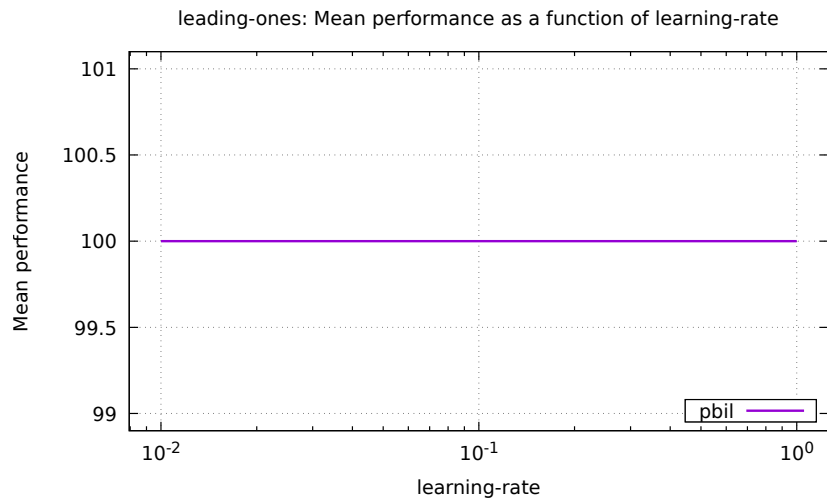
algorithm	learning-rate	min	$Q_1$	med.	$Q_3$	max	rk
pbil	0.01	100	100	100	100	100	1
pbil	0.02	100	100	100	100	100	1
pbil	0.05	100	100	100	100	100	1
pbil	0.1	100	100	100	100	100	1
pbil	0.2	100	100	100	100	100	1
pbil	0.5	100	100	100	100	100	1
pbil	1	100	100	100	100	100	1

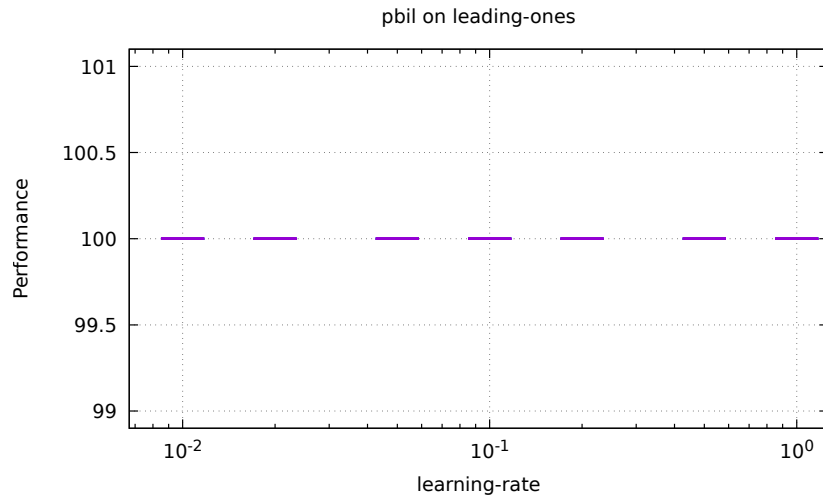
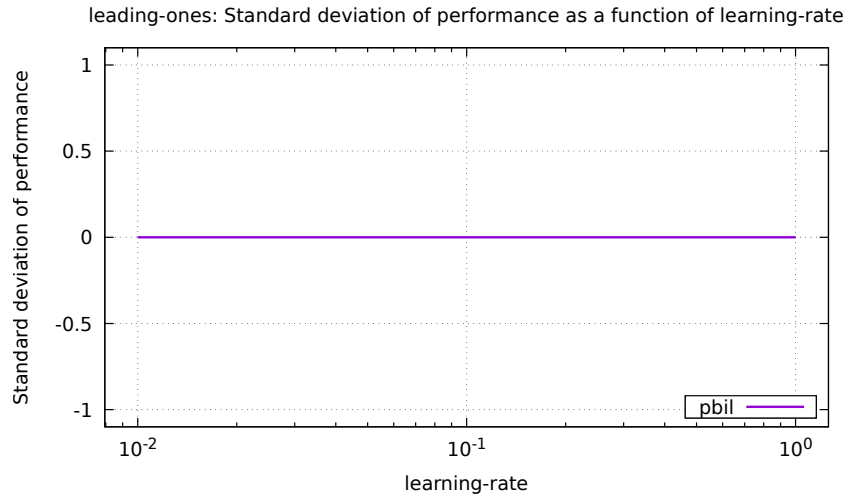




## 4 Function leading-ones

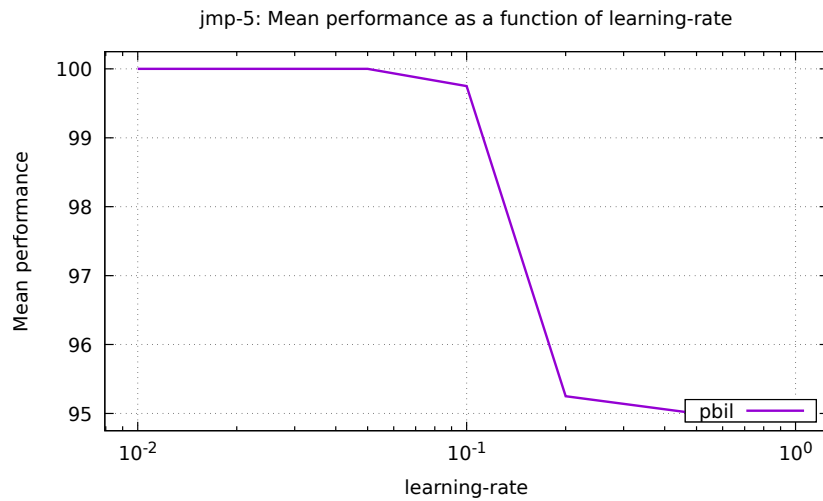
algorithm	learning-rate	min	$Q_1$	med.	$Q_3$	max	rk
pbil	0.01	100	100	100	100	100	1
pbil	0.02	100	100	100	100	100	1
pbil	0.05	100	100	100	100	100	1
pbil	0.1	100	100	100	100	100	1
pbil	0.2	100	100	100	100	100	1
pbil	0.5	100	100	100	100	100	1
pbil	1	100	100	100	100	100	1

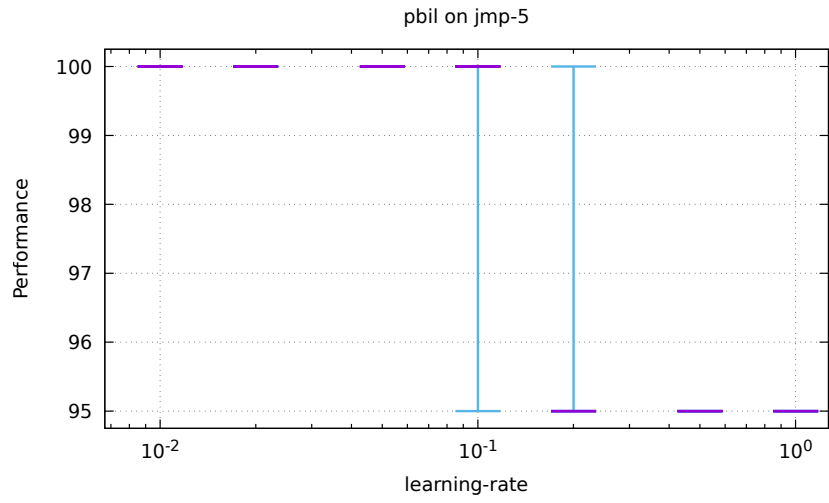
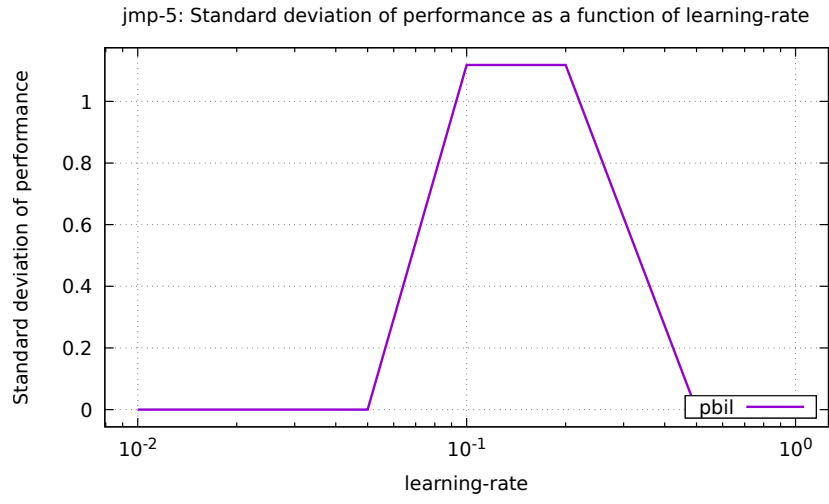




## 5 Function jmp-5

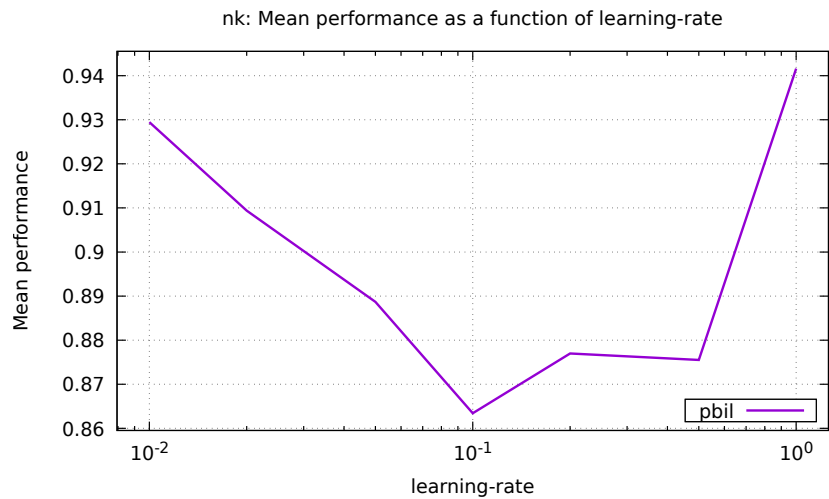
algorithm	learning-rate	min	$Q_1$	med.	$Q_3$	max	rk
pbil	0.01	100	100	100	100	100	1
pbil	0.02	100	100	100	100	100	1
pbil	0.05	100	100	100	100	100	1
pbil	0.1	95	100	100	100	100	4
pbil	0.2	95	95	95	95	100	5
pbil	0.5	95	95	95	95	95	6
pbil	1	95	95	95	95	95	6

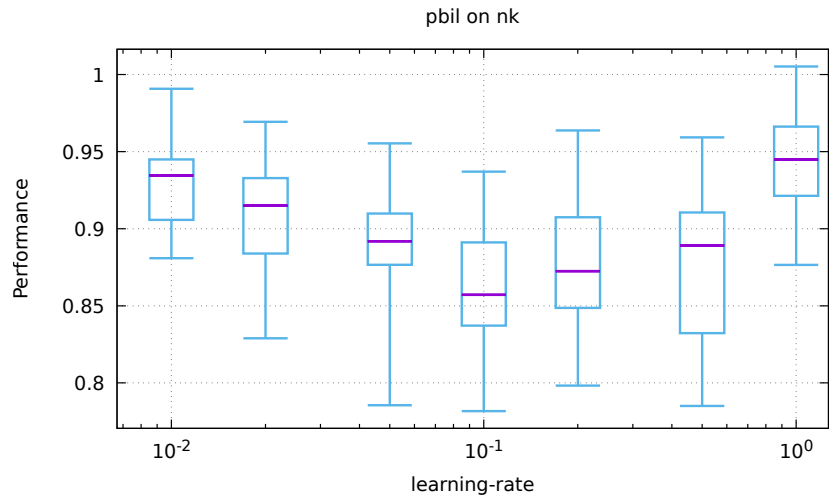
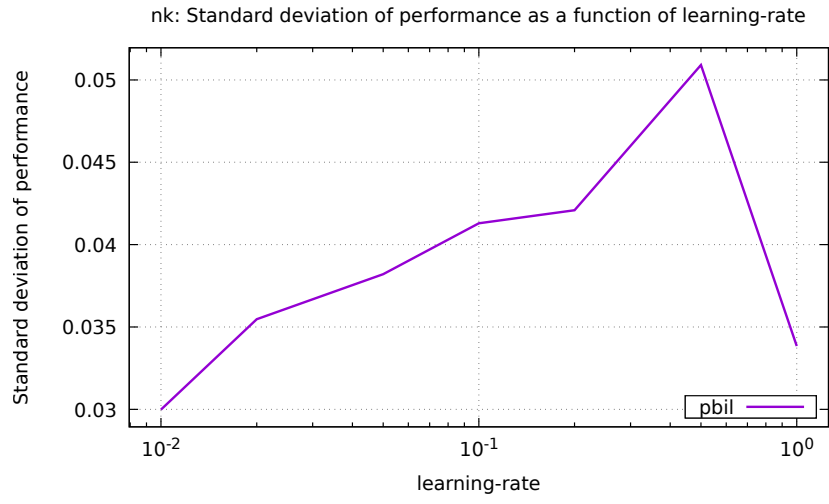




## 6 Function nk

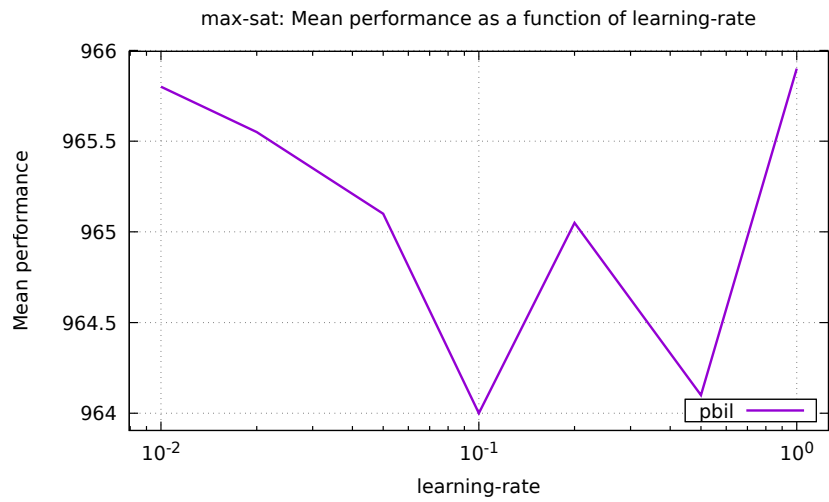
algorithm	learning-rate	min	$Q_1$	med.	$Q_3$	max	rk
pbil	0.01	0.88	0.91	0.93	0.94	0.99	2
pbil	0.02	0.83	0.88	0.91	0.93	0.97	3
pbil	0.05	0.79	0.88	0.89	0.91	0.96	4
pbil	0.1	0.78	0.84	0.86	0.89	0.94	7
pbil	0.2	0.80	0.85	0.87	0.91	0.96	6
pbil	0.5	0.79	0.83	0.89	0.91	0.96	5
pbil	1	0.88	0.92	0.94	0.97	1.01	1

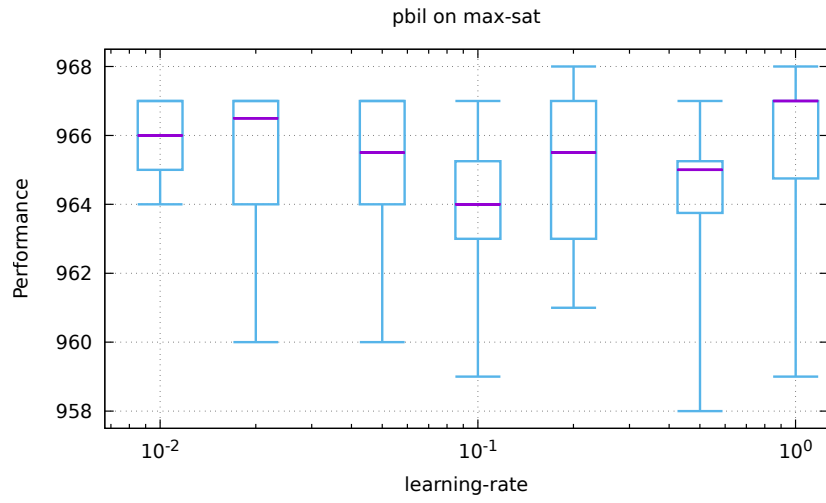
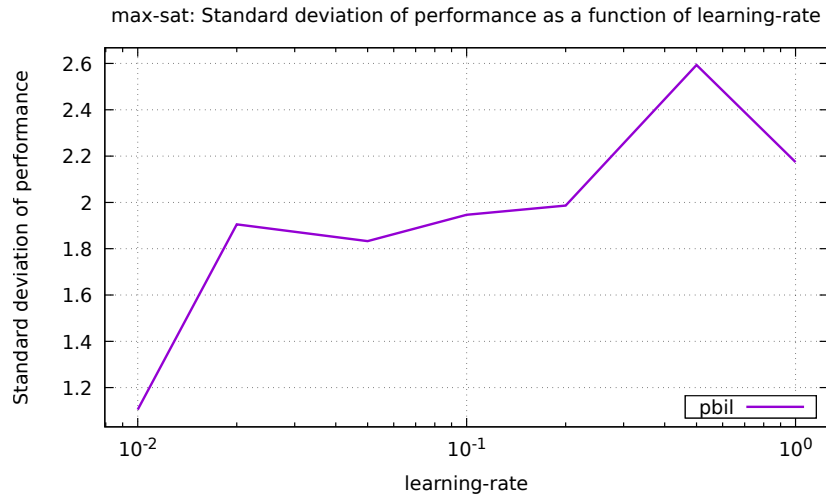




## 7 Function max-sat

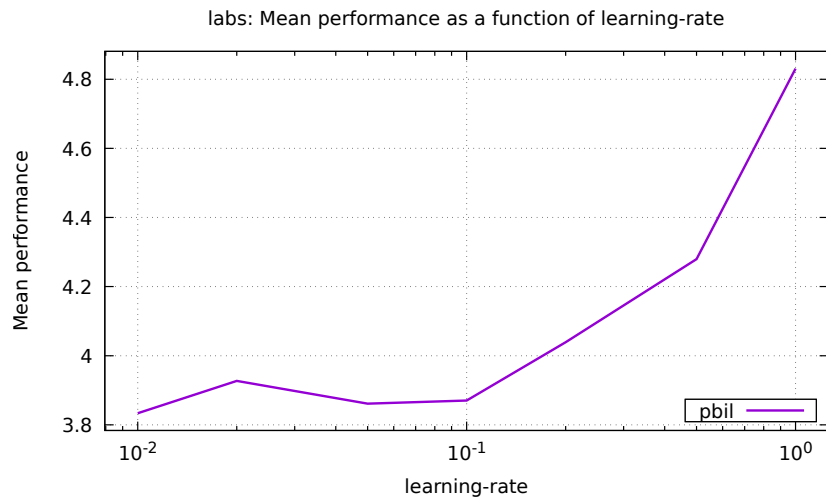
algorithm	learning-rate	min	$Q_1$	med.	$Q_3$	max	rk
pbil	0.01	964	965	966	967	967	3
pbil	0.02	960	964	967	967	967	2
pbil	0.05	960	964	966	967	967	4
pbil	0.1	959	963	964	965	967	7
pbil	0.2	961	963	966	967	968	5
pbil	0.5	958	964	965	965	967	6
pbil	1	959	965	967	967	968	1



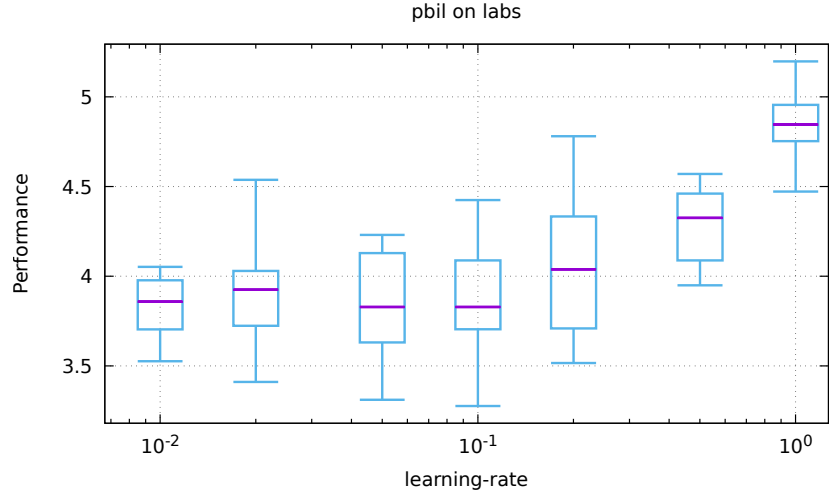
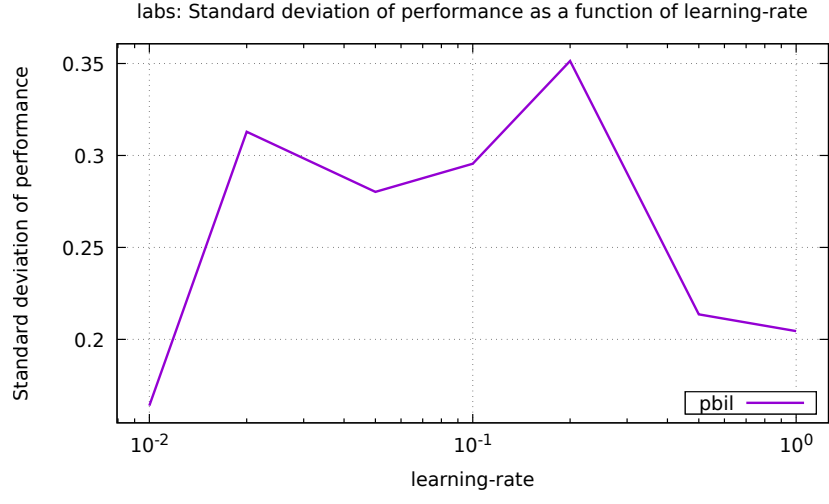


## 8 Function labs

algorithm	learning-rate	min	$Q_1$	med.	$Q_3$	max	rk
pbil	0.01	3.53	3.70	3.86	3.98	4.05	5
pbil	0.02	3.41	3.72	3.92	4.03	4.54	4
pbil	0.05	3.31	3.63	3.83	4.13	4.23	6
pbil	0.1	3.28	3.70	3.83	4.09	4.42	7
pbil	0.2	3.52	3.71	4.04	4.33	4.78	3
pbil	0.5	3.95	4.09	4.33	4.46	4.57	2
pbil	1	4.47	4.75	4.84	4.96	5.20	1

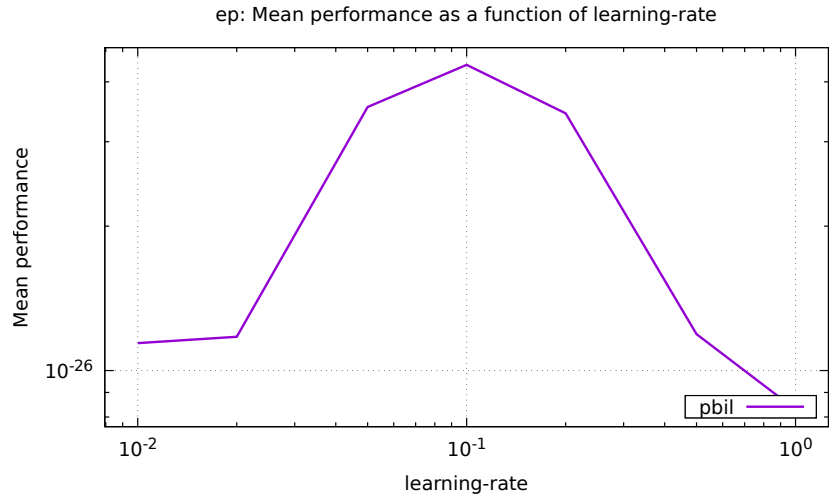


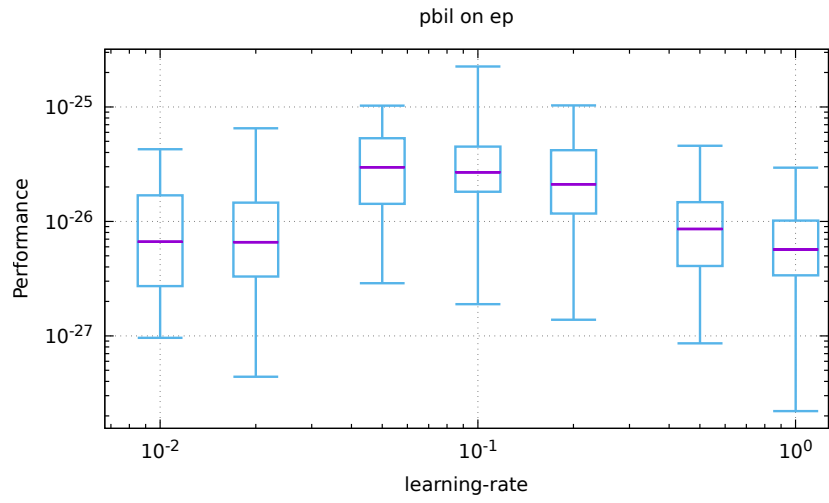
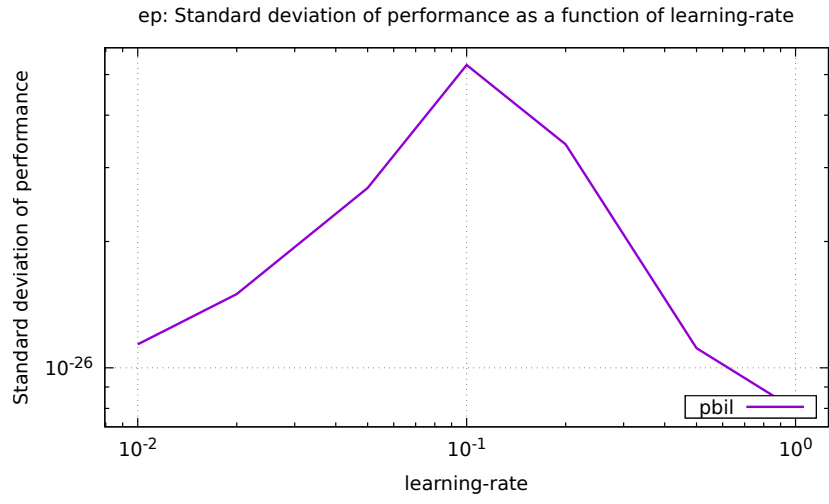




## 9 Function ep

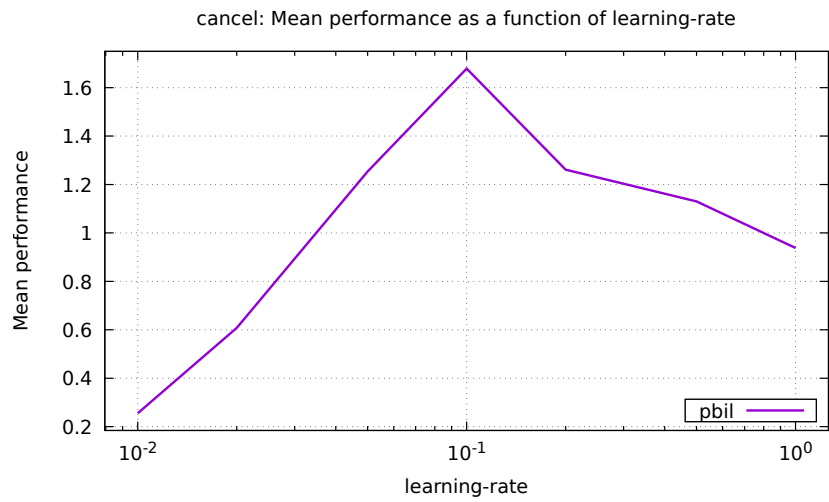
algorithm	learning-rate	min	$Q_1$	med.	$Q_3$	max	rk
pbil	0.01	$9.6 \times 10^{-28}$	$2.7 \times 10^{-27}$	$6.6 \times 10^{-27}$	$1.7 \times 10^{-26}$	$4.3 \times 10^{-26}$	3
pbil	0.02	$4.4 \times 10^{-28}$	$3.3 \times 10^{-27}$	$6.6 \times 10^{-27}$	$1.5 \times 10^{-26}$	$6.5 \times 10^{-26}$	2
pbil	0.05	$2.9 \times 10^{-27}$	$1.4 \times 10^{-26}$	$3.0 \times 10^{-26}$	$5.3 \times 10^{-26}$	$1.0 \times 10^{-25}$	7
pbil	0.1	$1.9 \times 10^{-27}$	$1.8 \times 10^{-26}$	$2.7 \times 10^{-26}$	$4.5 \times 10^{-26}$	$2.3 \times 10^{-25}$	6
pbil	0.2	$1.4 \times 10^{-27}$	$1.2 \times 10^{-26}$	$2.1 \times 10^{-26}$	$4.2 \times 10^{-26}$	$1.0 \times 10^{-25}$	5
pbil	0.5	$8.6 \times 10^{-28}$	$4.1 \times 10^{-27}$	$8.6 \times 10^{-27}$	$1.5 \times 10^{-26}$	$4.6 \times 10^{-26}$	4
pbil	1	$2.2 \times 10^{-28}$	$3.4 \times 10^{-27}$	$5.7 \times 10^{-27}$	$1.0 \times 10^{-26}$	$2.9 \times 10^{-26}$	1

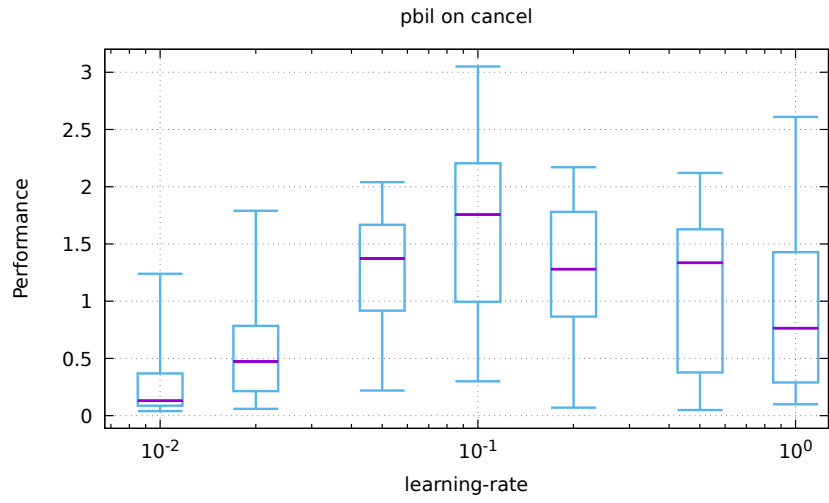
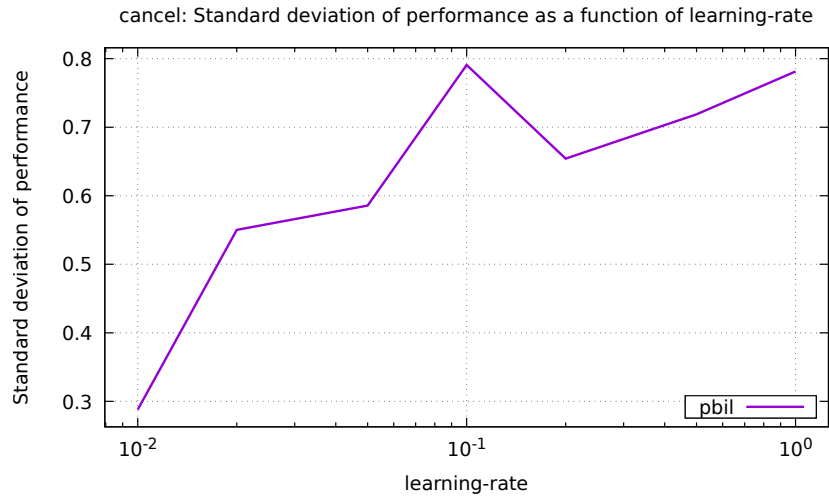




## 10 Function cancel

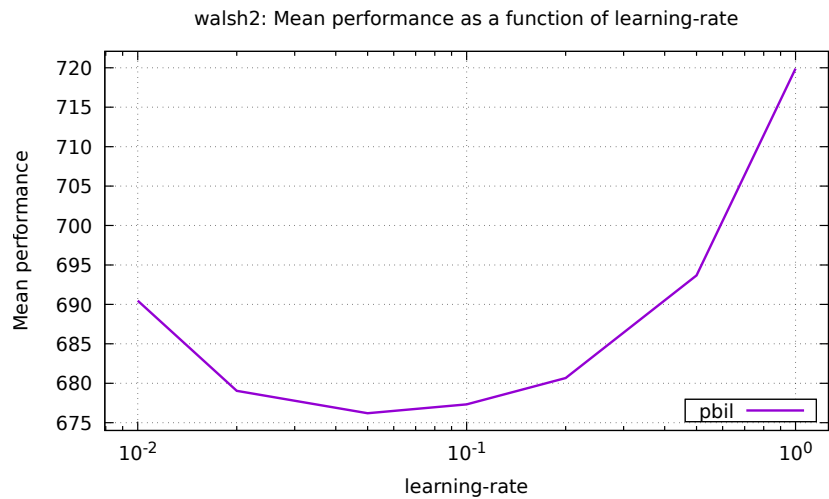
algorithm	learning-rate	min	$Q_1$	med.	$Q_3$	max	rk
pbil	0.01	0.04	0.09	0.13	0.37	1.24	1
pbil	0.02	0.06	0.22	0.48	0.79	1.79	2
pbil	0.05	0.22	0.92	1.38	1.67	2.04	6
pbil	0.1	0.30	1.00	1.76	2.21	3.05	7
pbil	0.2	0.07	0.87	1.28	1.78	2.17	4
pbil	0.5	0.05	0.38	1.34	1.63	2.12	5
pbil	1	0.10	0.29	0.77	1.43	2.61	3

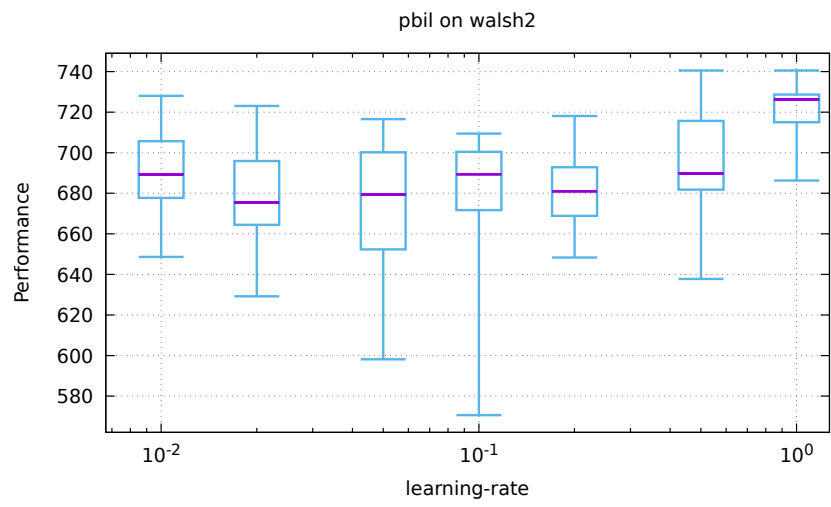
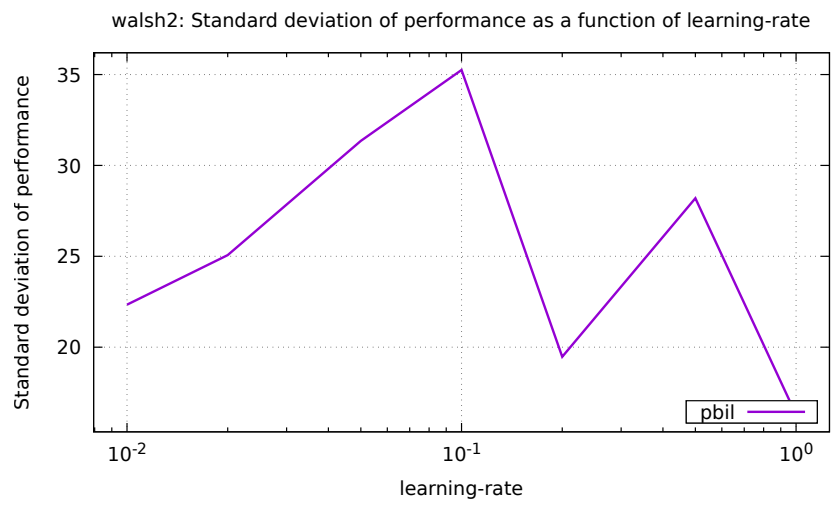




## 11 Function walsh2

algorithm	learning-rate	min	$Q_1$	med.	$Q_3$	max	rk
pbil	0.01	648.64	677.70	689.32	705.68	728.02	4
pbil	0.02	629.21	664.46	675.39	695.88	723.07	7
pbil	0.05	598.13	652.34	679.31	700.21	716.59	6
pbil	0.1	570.63	671.68	689.40	700.47	709.41	3
pbil	0.2	648.31	668.88	681.02	692.83	718.10	5
pbil	0.5	637.76	681.81	689.83	715.67	740.55	2
pbil	1	686.26	715.03	726.23	728.68	740.55	1





## 12 Default parameters

```
# algorithm = 100
# bm_mc_reset_strategy = 1
# bm_num_gs_cycles = 1
# bm_num_gs_steps = 100
# bm_sampling = 1
# budget = 10000
# bv_size = 100
# ea_lambda = 100
# ea_mu = 10
# fun_name = noname
# fun_num_traps = 10
# fun_prefix_length = 2
# fun_threshold = 10
# function = 0
# ga_crossover_probability = 0.5
# ga_tournament_size = 10
# hea_binary_dynamics = 0
# hea_delay = 10000
# hea_num_par_updates = 1
# hea_num_seq_updates = 100
# hea_rate_strategy = 0
# hea_reset_period = 0
# hea_sampling_method = 0
# hea_time_constant = 1000
# hea_weight = 1
# learning_rate = 0.001
# map = 0
# map_input_size = 100
# map_path = nopath
# neighborhood = 0
# neighborhood_iterator = 0
# noise_stddev = 1
# num_iterations = 0
# num_threads = 1
# path = nopath
# population_size = 10
# pv_log_num_components = 5
# radius = 2
# rls_patience = 50
# sa_initial_acceptance_probability = 0.6
# sa_num_transitions = 50
# sa_num_trials = 100
# sa_rate = 1.2
# scaled_mutation_probability = 1
# seed = 0
# selection_size = 1
# target = 100
# print_default_parameters
# last_parameter
# exec_name = hnco
# version = 0.7
# Generated from hnco.json
```