Improving performance using scipy.sparse

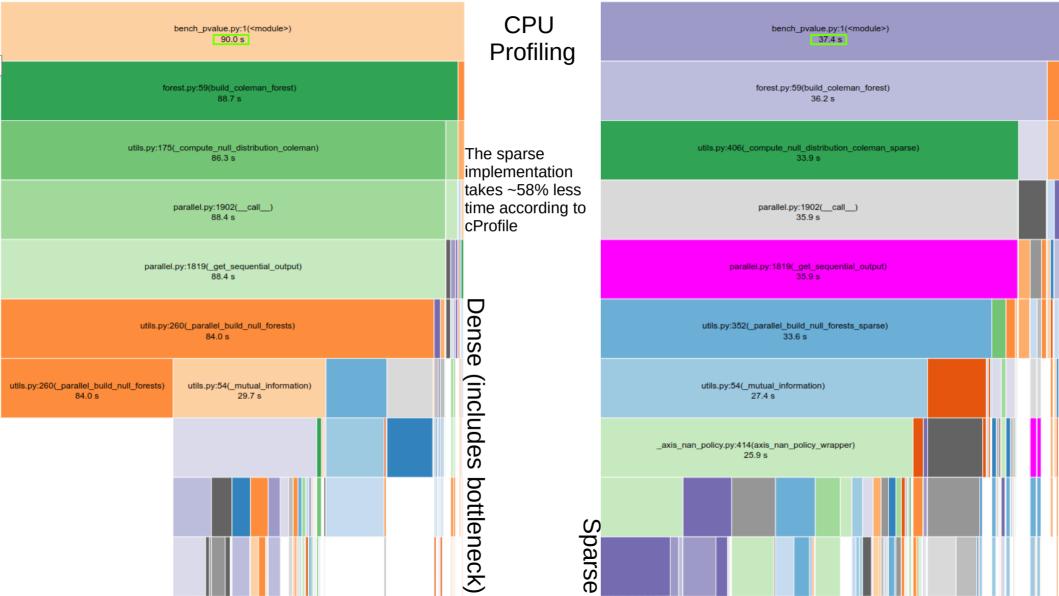
Note that this only works in the special case of binary classification or regression as scipy.sparse only supports 2d matrices.

This could be generalized using pydata.sparse but currently that uses too much memory.

The next two slides show the CPU and Memory profiling analysis of the script to the right.

The profilers seem to indicate that using scipy.sparse cuts runtime by more than half and uses 7% less memory.

nport sys treeple.datasets import make_trunk_classification treeple.ensemble import HonestForestClassifier treeple.stats import PermutationHonestForestClassifier, build_coleman_forest N PERMUTATIONS = 10000y = make_trunk_classification(n_samples=1000. n_dim=1, mu_0=0, mu 1=1 n_informative=1. seed=1, est = HonestForestClassifier(n_estimators=100, max samples=1.6. max_features=0.3 bootstrap=True, random_state=1. est_null = PermutationHonestForestClassifier(n_estimators=100, max_samples=1.6 max_features=0.3 bootstrap=True, random_state=1, observed_diff, _, _, pvalue, mix_diff = build_coleman_forest(est_null, metric="mi" n_repeats=N_PERMUTATIONS, return_posteriors=False, seed=1. use_sparse="sparse" in sys.argv,



Memory Profiling using memray

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runpy.run_module(args.script, run_name="main", alter_sys=True)											
observed_diff, _, _, pvalue, mix_diff = build_coleman_forest(
est, orig_forest_proba = buil metric_star, metric_star_pi = _compute_nu					_compute_null_distr	ribution_coleman(perm_est, perm_forest_proba				
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Dense

Total for benchmarking script: ~129MB Total for build_coleman_forest: ~24.2MB

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est, orig_forest_proba = build_oob perm_est, perm_forest_proba = build_						oocompute_null_distribution_coleman_sparse(
est.fit(X, y.ravel(), **est_kwargs)			est.fit(X, y.ravel(), **est_kwargs)				out = Parallel(n_job		for _, seed in zip(range(n_repe	all		
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	builder.build(self.tree			self.estimatorfit(
				self = selfbuild_tree(
				builder.build(self.tree_, X, y,								

Sparse

Total for benchmarking script: ~120.1MB Total for build_coleman_forest: ~15.2MB