Overview

This package provides functions to estimate a skeleton graph and a completed partially acyclic graph (CPDAG) from a data matrix that describes the appearance patterns of the graph nodes. The detailed algorithm is written in [Kalisch2007].

Code

The source code is available at https://github.com/keiichishima/pcalg

Bug Reports

GPUCSL

GPUCSL is a python library for constraint-based causal structure learning using Graphics Processing Units (GPUs). In particular, it utilizes CUDA for GPU acceleration to speed up the well-known PC algorithm.

Features

- Performant GPU implementation of the PC algorithm (covers discrete, and multivariate normal distributed data),
 see General Notes;
- Multi-GPU support (experimental; for now, only for gaussian CI kernel);
- Easy to install and use, see Usage;
- A Command Line Interface (CLI);

- Took only 18 seconds to read in the entire first chromosome and generate a CPDAG (31 column x 4644 row dataset)
- Using Google Colab's Tesla T4 GPU with 16GB of memory.

Result: DiGraph with 31 nodes and 73 edges

- [(0, 4), (0, 8), (0, 12), (0, 26), (0, 30), (1, 3), (1, 7), (1, 15), (1, 19), (1, 23), (1, 27), (2, 6), (2, 10), (2, 16), (2, 18), (2, 26), (3, 7), (3, 9), (3, 11), (3, 15), (3, 19), (3, 23), (3, 27), (4, 8), (4, 10), (4, 16), (4, 28), (4, 30), (5, 9), (5, 19), (5, 23), (5, 27), (6, 12), (6, 22), (7, 11), (7, 27), (8, 16), (8, 20), (8, 26), (9, 11), (9, 19), (9, 23), (9, 27), (10, 18), (10, 22), (11, 15), (11, 19), (11, 23), (11, 27), (12, 14), (12, 16), (12, 20), (13, 19), (13, 23), (13, 27), (14, 18), (14, 20), (14, 22), (15, 19), (15, 27), (16, 24), (16, 28), (17, 23), (18, 22), (20, 24), (20, 26), (20, 28), (21, 23), (22, 26), (23, 27), (25, 27), (26, 30), (28, 30)]
- *shortened to column index. Still have access to nodes corresponding to names

- Wanted to build off last week's hypothesis
- Unable to figure out how to one hot encode a column of a ndarray (python)
- Chose to do all preprocessing in R and save the .csv
- No longer works

(new non-working array) (old working array) class: ndarray class: ndarray shape: (4644, 4658) shape: (4644, 31) strides: (37264, 8) strides: (248, 8) itemsize: 8 itemsize: 8 aligned: True aligned: True contiguous: True contiguous: True fortran: False fortran: False data pointer: 0x583a7dbdc2e0 data pointer: 0x5c5677b39a20 byteorder: little byteorder: little byteswap: False byteswap: False type: float64 type: float64

```
/usr/local/lib/python3.10/dist-packages/gpucsl/pc/gaussian device manager.py in compute_skeleton(self,
                    final skeleton = self.execute ci workers in parallel(worker function)
    188
    189
                else:
--> 190
                    worker_function(0, self)
   191
                    final_skeleton = self.get_skeleton_for_device_index(0).get()
   192
/usr/local/lib/python3.10/dist-packages/gpucsl/pc/discover skeleton gaussian.py in gaussian_ci_worker(d
    96
    97
---> 98
                        stream.synchronize()
    99
                        log time(
                            f"[T: {device index} D: {device}] Level {level} time".
    100
cupy/cuda/stream.pyx in cupy.cuda.stream. BaseStream.synchronize()
cupy backends/cuda/api/runtime.pyx in cupy backends.cuda.api.runtime.streamSynchronize()
cupy_backends/cuda/api/runtime.pyx in cupy_backends.cuda.api.runtime.check_status()
CUDARuntimeError: cudaErrorIllegalAddress: an illegal memory access was encountered
```

---> 34

35 36 result = f(*args, **kw)

duration = end - start

end = timer()