

In [15]: `pip install rpy2`

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Requirement already satisfied: rpy2 in /opt/anaconda3/lib/python3.13/site-packages (3.6.4)
Requirement already satisfied: rpy2-rinterf>=3.6.3 in /opt/anaconda3/lib/python3.13/site-packages (from rpy2) (3.6.3)
Requirement already satisfied: rpy2-robjects>=3.6.3 in /opt/anaconda3/lib/python3.13/site-packages (from rpy2) (3.6.3)
Requirement already satisfied: cffi>=1.15.1 in /opt/anaconda3/lib/python3.13/site-packages (from rpy2-rinterf>=3.6.3->rpy2) (1.17.1)
Requirement already satisfied: pycparser in /opt/anaconda3/lib/python3.13/site-packages (from cffi>=1.15.1->rpy2-rinterf>=3.6.3->rpy2) (2.21)
Requirement already satisfied: jinja2 in /opt/anaconda3/lib/python3.13/site-packages (from rpy2-robjects>=3.6.3->rpy2) (3.1.6)
Requirement already satisfied: tzlocal in /opt/anaconda3/lib/python3.13/site-packages (from rpy2-robjects>=3.6.3->rpy2) (5.3.1)
Requirement already satisfied: MarkupSafe>=2.0 in /opt/anaconda3/lib/python3.13/site-packages (from jinja2->rpy2-robjects>=3.6.3->rpy2) (3.0.2)
Note: you may need to restart the kernel to use updated packages.
```

In [20]: `import os`  
`os.environ["R_HOME"] = "/Library/Frameworks/R.framework/Resources"`

In [21]: `print("R_HOME =", os.environ.get("R_HOME"))`

R\_HOME = /Library/Frameworks/R.framework/Resources

In [22]: `import numpy as np`  
`import itertools`  
`from math import log, sqrt`  
`from collections import defaultdict`  
`from dataclasses import dataclass`  
  
`# If you don't have SciPy, replace this with an approximation`  
`from scipy.stats import norm`

```
# -----
# 1. Conditional independence test: Fisher-Z for Gaussian data
# -----
```

```
def fisher_z_test(Z, i, j, cond, alpha):
    """
    Gaussian CI test using Fisher-Z.
    Z: data matrix (n_samples, n_nodes)
    i,j: indices of variables
    cond: iterable of indices (conditioning set)
    alpha: significance level
    Returns: True if  $X_i \perp X_j \mid \text{cond}$  (independent)
    """
    n, p = Z.shape
    var_idx = [i, j] + list(cond)
    sub = Z[:, var_idx]
    C = np.cov(sub, rowvar=False)
    # precision
```

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try:
    K = np.linalg.inv(C)
except np.linalg.LinAlgError:
    # fall back to pseudo-inverse
    K = np.linalg.pinv(C)
# partial correlation between first two
r = -K[0, 1] / np.sqrt(K[0, 0] * K[1, 1])
r = max(min(r, 0.999999), -0.999999)
z = 0.5 * log((1 + r) / (1 - r)) * sqrt(max(n - len(cond) - 3, 1))
zcrit = norm.ppf(1 - alpha / 2.0)
return abs(z) <= zcrit

# -----
# 2. Dynamic PAG graph with homology (SVAR structure)
# -----

@dataclass
class NodeInfo:
    var: int
    lag: int

class DynamicPAG:
    """
    Dynamic PAG segment for X_t, ..., X_{t-p}.
    Nodes are indexed 0..(k*(p+1)-1).
    Each node has (var, lag). lag=0 is time t (most recent).
    marks[i,j] is the endpoint mark at j for edge i--j:
        0 = circle (o)
        -1 = tail (-)
        1 = arrow (>)
    """
    def __init__(self, var_names, max_lag):
        self.var_names = list(var_names)
        self.k = len(var_names)
        self.p = max_lag
        self.n_nodes = self.k * (self.p + 1)

        # full complete graph with o-o edges
        self.adj = np.ones((self.n_nodes, self.n_nodes), dtype=bool)
        np.fill_diagonal(self.adj, False)
        self.marks = np.zeros((self.n_nodes, self.n_nodes), dtype=int)

        # separation sets: key = (min(i,j), max(i,j)), value = set of nodes
        self.sepset = {}

    # ----- node indexing / decoding -----

    def node_index(self, var, lag):
        return lag * self.k + var

    def decode_node(self, idx):
        lag = idx // self.k
        var = idx % self.k
        return NodeInfo(var=var, lag=lag)

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def node_label(self, idx):
    info = self.decode_node(idx)
    return f"{self.var_names[info.var]}_lag{info.lag}"

# ----- homology: pairs with same var pair + same lag difference -----

def hom_pairs(self, i, j):
    info_i = self.decode_node(i)
    info_j = self.decode_node(j)
    d = info_i.lag - info_j.lag
    pairs = []
    for a in range(self.p + 1):
        b = a - d
        if 0 <= b <= self.p:
            m = self.node_index(info_i.var, a)
            n = self.node_index(info_j.var, b)
            pairs.append((m, n))
    return pairs

# ----- adjacency restricted by time (adj_t) -----

def neighbors(self, i):
    return [j for j in range(self.n_nodes) if self.adj[i, j]]

def adj_t(self, i):
    info_i = self.decode_node(i)
    res = []
    for j in self.neighbors(i):
        info_j = self.decode_node(j)
        if info_j.lag >= info_i.lag:
            res.append(j)
    return res

# ----- sepsets -----

def set_sepset(self, i, j, S):
    if i > j:
        i, j = j, i
    self.sepset[(i, j)] = set(S)

def get_sepset(self, i, j):
    if i > j:
        i, j = j, i
    return self.sepset.get((i, j), set())

# ----- edge operations with homology -----

def delete_edge_with_homology(self, i, j):
    for m, n in self.hom_pairs(i, j):
        if self.adj[m, n]:
            self.adj[m, n] = False
            self.adj[n, m] = False
            self.marks[m, n] = 0
            self.marks[n, m] = 0

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def _orient_edge(self, i, j, mark_ij, mark_ji):
    """
    Set endpoint marks for edge i-j without touching homology.
    mark_ij is mark at j on edge from i to j.
    """
    if not self.adj[i, j]:
        return
    # simple consistency check: don't overwrite a hard arrow in opposite
    if self.marks[j, i] == 1 and mark_ij == 1:
        # would make i <-> j when there is already arrow at i
        pass
    self.marks[i, j] = mark_ij
    self.marks[j, i] = mark_ji

def orient_with_homology(self, i, j, mark_ij, mark_ji):
    """
    Orient edge (i,j) and all homologous edges with same endpoint pattern
    For example, for i *-> j we pass (1, -1) or (1, 0), etc.
    """
    for m, n in self.hom_pairs(i, j):
        if not self.adj[m, n]:
            continue
        self._orient_edge(m, n, mark_ij, mark_ji)

def reset_all_to_oo(self):
    """Keep adjacency but set all marks to circle (o-o)."""
    self.marks[:, :] = 0

# ----- helpers for collider / triangle checks -----

def is_collider(self, a, b, c):
    """
    a ?-> b <-? c
    """
    if not (self.adj[a, b] and self.adj[b, c]):
        return False
    return self.marks[a, b] == 1 and self.marks[c, b] == 1

def forms_triangle(self, a, b, c):
    return self.adj[a, b] and self.adj[b, c] and self.adj[a, c]

# -----
# 3. Dynamic pds_s (time-restricted possible-d-sep)
# -----

def pds_s(graph: DynamicPAG, i, j):
    """
    Time-restricted possible-d-sep set pds_s(X_i, X_j, P).
    This is a faithful adaptation of Zhang's pds definition,
    restricted to nodes whose lag <= max(lag(i), lag(j)).
    """
    info_i = graph.decode_node(i)
    info_j = graph.decode_node(j)
    maxlag = max(info_i.lag, info_j.lag)

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# BFS over paths satisfying collider-or-triangle condition
result = set()
queue = [[i]]
visited_paths = set()

def path_ok(path):
    # all internal triples (a,b,c) must be collider or triangle
    if len(path) < 3:
        return True
    for a, b, c in zip(path[:-2], path[1:-1], path[2:]):
        if not (graph.is_collider(a, b, c) or graph.forms_triangle(a, b, c)):
            return False
    return True

while queue:
    path = queue.pop(0)
    last = path[-1]
    for nb in graph.neighbors(last):
        if nb in path:
            continue
        new_path = path + [nb]
        key = tuple(new_path)
        if key in visited_paths:
            continue
        visited_paths.add(key)
        if not path_ok(new_path):
            continue

        info_nb = graph.decode_node(nb)
        if info_nb.lag <= maxlag:
            result.add(nb)
            queue.append(new_path)

result.discard(i)
result.discard(j)
return result

# -----
# 4. SVAR-FCI Algorithm 3.1
# -----

class SVAR_FCI:
    def __init__(self, alpha=0.05, max_lag=2, verbose=False):
        self.alpha = alpha
        self.max_lag = max_lag
        self.verbose = verbose
        self.graph_ = None
        self.var_names_ = None
        self.Z_ = None # lagged data

    # --- lagged data builder (X_t,...,X_{t-p}) ---

    def _build_lagged_matrix(self, X, var_names):
        """
        X: np.ndarray (T, k)

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Returns Z: (T-p, k*(p+1)), names: list[str]
"""
T, k = X.shape
p = self.max_lag
rows = T - p
Z = np.zeros((rows, k * (p + 1)))
names = []
for lag in range(p + 1):
    Z[:, lag * k:(lag + 1) * k] = X[p - lag:T - lag, :]
    for idx, name in enumerate(var_names):
        names.append(f"{name}_lag{lag}")
return Z, names

# --- independence wrapper ---

def _indep(self, Z, i, j, S):
    return fisher_z_test(Z, i, j, S, self.alpha)

# --- skeleton phase (Alg 3.1 lines 3-8) ---

def _skeleton_phase(self, G: DynamicPAG, Z):
    if self.verbose:
        print("Skeleton phase (dynamic adj_t + homology)...")
    n = 0
    p = G.n_nodes
    changed = True
    while changed:
        changed = False
        if self.verbose:
            print(f"Conditioning set size n={n}")
        for i in range(p):
            for j in range(i + 1, p):
                if not G.adj[i, j]:
                    continue
                # we test only with Xi as "left" node (as in Alg 3.1)
                adj_i_t = [v for v in G.adj_t(i) if v != j]
                if len(adj_i_t) < n:
                    continue
                found_sep = False
                for S in itertools.combinations(adj_i_t, n):
                    if self._indep(Z, i, j, S):
                        if self.verbose:
                            print(f"    indep({G.node_label(i)}, {G.node_label(j)} | {S})")
                        G.delete_edge_with_homology(i, j)
                        G.set_sepset(i, j, S)
                        changed = True
                        found_sep = True
                        break
                if found_sep:
                    continue
            n += 1

# --- line 9: time orientation Xi_t *-> Xj_s if s > t ---

# --- line 9: time orientation Xi_t *-> Xj_s if s > t (past -> future) ---

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def _time_orientation(self, G: DynamicPAG):
    """
    Deterministic time-based orientation:

    We use lag = 0 for time t (most recent), lag = 1 for t-1, etc.
    So a larger lag means an *earlier* time.

    The SVAR-FCI rule  $X_{i,t} \rightarrow X_{j,s}$  iff  $s > t$  (later in calendar time)
    therefore translates to:

        if lag(i) > lag(j): # i is further in the past than j
            orient  $X_{i,lag(i)} \rightarrow X_{j,lag(j)}$ 

    i.e. edges always point from past  $\rightarrow$  future.
    """
    if self.verbose:
        print("Time orientation (past  $\rightarrow$  future: larger lag  $\rightarrow$  smaller lag)")
    n = G.n_nodes
    for i in range(n):
        info_i = G.decode_node(i)
        for j in range(n):
            if not G.adj[i, j]:
                continue
            info_j = G.decode_node(j)

            # i earlier in time than j  $\Leftrightarrow$  lag(i) > lag(j)
            if info_i.lag > info_j.lag:
                # orient i  $\rightarrow$  j (circle at i, arrow at j), with homology
                if G.marks[i, j] == 0 and G.marks[j, i] == 0:
                    G.orient_with_homology(i, j, 1, 0)

# --- line 10: v-structures with homology ---

def _orient_v_structures(self, G: DynamicPAG):
    if self.verbose:
        print("Orienting v-structures...")
    p = G.n_nodes
    for k in range(p):
        for i in range(p):
            if i == k or not G.adj[i, k]:
                continue
            for j in range(i + 1, p):
                if j == k or not G.adj[j, k]:
                    continue
                if G.adj[i, j]:
                    continue # shielded
                S = G.get_sepset(i, j)
                if k not in S:
                    # orient i  $\rightarrow$  k  $\leftarrow$  j
                    G.orient_with_homology(i, k, 1, -1)
                    G.orient_with_homology(j, k, 1, -1)

# --- line 11: second deletion using dynamic pds_s + homology ---

def _pds_deletion_phase(self, G: DynamicPAG, Z):

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if self.verbose:
    print("pds_s deletion phase with homology...")
n = 0
p = G.n_nodes
changed = True
while changed:
    changed = False
    if self.verbose:
        print(f"  pds_s, conditioning size n={n}")
    for i in range(p):
        for j in range(i + 1, p):
            if not G.adj[i, j]:
                continue
            # candidate conditioning sets from pds_s
            P1 = pds_s(G, i, j)
            P2 = pds_s(G, j, i)
            P_union = list(P1.union(P2))
            if len(P_union) < n:
                continue
            found_sep = False
            for S in itertools.combinations(P_union, n):
                if self._indep(Z, i, j, S):
                    if self.verbose:
                        print(f"    pds indep({G.node_label(i)}, {G.
G.delete_edge_with_homology(i, j)
G.set_sepset(i, j, S)
changed = True
found_sep = True
                    break
            if found_sep:
                continue
        n += 1

# --- R1-R10 placeholder (needs full Zhang implementation) ---

# --- R1-R10 (Zhang 2008) orientation rules on DynamicPAG ---
# Assumptions:
# - G.marks[i,j] is mark at endpoint j on edge (i,j):
#     0 = circle (o)
#    -1 = tail   (-)
#     1 = arrow  (>)
# - G.adj[i,j] is True iff there is an edge between i and j.
# - G.orient_with_homology(i,j, mark_ij, mark_ji) orients (i,j) and
#   all homologous edges with endpoint marks mark_ij at j, mark_ji at i.
#
# We implement R1-R4 and R8-R10 exactly in Zhang (2008), restricted
# to MAGs without undirected edges (no selection bias), so R5-R7
# are omitted as per Zhang's remark.

def _apply_R_rules(self, G: DynamicPAG):
    n = G.n_nodes

    def non_adjacent(a, b):
        return not G.adj[a, b]

    def is_arrow_into(child, parent):

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# edge parent *-> child ⇔ mark at child on (parent, child) is 1
return G.adj[parent, child] and G.marks[parent, child] == 1

def is_circle_at(i, j):
    # circle at j on edge (i,j)
    return G.adj[i, j] and G.marks[i, j] == 0

# --- R4 helpers: discriminating paths ---

def is_discriminating_path(path, V, X, Y):
    """
    Check if 'path' (list of nodes [X,...,W,V,Y]) is a
    discriminating path for V between X and Y, in the sense of
    Zhang Def. 7 but adapted to the current PAG:
    - length >= 3 edges (>= 4 nodes)
    - V non-endpoint and adjacent to Y on path
    - X not adjacent to Y
    - every vertex between X and V is:
      * a collider on the path
      * a parent of Y (edge *-> Y).
    """
    if len(path) < 4:
        return False
    if path[0] != X or path[-1] != Y:
        return False
    if V not in path[1:-1]:
        return False
    if not non_adjacent(X, Y):
        return False

    V_idx = path.index(V)
    # V must be non-endpoint and directly before Y on the path
    if V_idx == 0 or V_idx == len(path) - 1:
        return False
    if path[V_idx + 1] != Y:
        return False

    # Every vertex between X and V must be a collider on the path
    # and must be a parent of Y (*-> Y).
    for idx in range(1, V_idx):
        v = path[idx]
        prev_v = path[idx - 1]
        next_v = path[idx + 1]
        # collider on the path
        if not G.is_collider(prev_v, v, next_v):
            return False
        # parent of Y: edge v *-> Y ⇒ mark at Y on (v,Y) is 1
        if not is_arrow_into(Y, v):
            return False

    return True

def all_simple_paths_unbounded(start, end):
    """
    Generate all simple paths from start to end (no repeated nodes).
    Use with care; graph is sparse after skeleton + pds, so it is

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manageable for typical macro VAR sizes.
"""
stack = [(start, [start])]
while stack:
    (v, path) = stack.pop()
    for w in G.neighbors(v):
        if w in path:
            continue
        new_path = path + [w]
        if w == end:
            yield new_path
        else:
            stack.append((w, new_path))

changed = True
while changed:
    changed = False

    # ----- R1 -----
    # R1: If  $\alpha \rightarrow \beta \rightarrow \gamma$  and  $\alpha, \gamma$  nonadjacent, then orient  $\beta \rightarrow \gamma$ 
    for beta in range(n):
        for alpha in G.neighbors(beta):
            if not is_arrow_into(beta, alpha):
                continue # need  $\alpha \rightarrow \beta$ 
            for gamma in G.neighbors(beta):
                if gamma == alpha:
                    continue
                #  $\beta \rightarrow \gamma \rightarrow$  circle at  $\beta$  on edge  $(\beta, \gamma)$ 
                if not is_circle_at(beta, gamma):
                    continue
                if not non_adjacent(alpha, gamma):
                    continue
                # orient  $\beta \rightarrow \gamma$  (tail at  $\beta$ , arrow at  $\gamma$ )
                if not (G.marks[beta, gamma] == 1 and G.marks[gamma,
                    G.orient_with_homology(beta, gamma, 1, -1)
                    changed = True

    # ----- R2 -----
    # R2: If  $\alpha \rightarrow \beta \rightarrow \gamma$  (or  $\alpha \rightarrow \beta \rightarrow \gamma$ ) and  $\alpha \rightarrow \gamma$ , orient  $\alpha \rightarrow \gamma$ 
    for beta in range(n):
        for alpha in G.neighbors(beta):
            for gamma in G.neighbors(beta):
                if gamma == alpha:
                    continue
                # require  $\beta \rightarrow \gamma$  and  $\alpha \rightarrow \beta$  (arrowheads into  $\beta$  and
                if not (is_arrow_into(beta, alpha) and is_arrow_into
                    continue
                if not G.adj[alpha, gamma]:
                    continue
                #  $\alpha \rightarrow \gamma \Rightarrow$  circle at  $\gamma$  on  $(\alpha, \gamma)$ 
                if G.marks[gamma, alpha] != 0:
                    continue
                # orient  $\alpha \rightarrow \gamma$ : arrow at  $\gamma$ , keep circle at  $\alpha$ 
                if not (G.marks[alpha, gamma] == 1 and G.marks[gamma,
                    G.orient_with_homology(alpha, gamma, 1, 0)
                    changed = True

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# ----- R3 -----
# R3: If  $\alpha \ast \rightarrow \beta \leftarrow \ast \gamma$ ,  $\alpha \ast \rightarrow \theta \circ \ast \gamma$ ,  $\alpha, \gamma$  nonadjacent,
# and  $\theta \ast \rightarrow \beta$ , then orient  $\theta \ast \rightarrow \beta$  as  $\theta \ast \rightarrow \beta$ .
for beta in range(n):
    # collider  $\alpha \ast \rightarrow \beta \leftarrow \ast \gamma$ 
    for alpha in G.neighbors(beta):
        if not is_arrow_into(beta, alpha):
            continue
    for gamma in G.neighbors(beta):
        if gamma == alpha:
            continue
        if not is_arrow_into(beta, gamma):
            continue
        if not non_adjacent(alpha, gamma):
            continue
    # search  $\theta$  such that  $\alpha \ast \rightarrow \theta \circ \ast \gamma$  and  $\theta \ast \rightarrow \beta$ 
    for theta in range(n):
        if theta in (alpha, beta, gamma):
            continue
        if not (G.adj[alpha, theta] and G.adj[theta, gamma]):
            continue
        # circles at  $\theta$  on  $\alpha-\theta$  and  $\gamma-\theta$ 
        if G.marks[alpha, theta] != 0 or G.marks[gamma, theta] != 0:
            continue
        # circle at  $\beta$  on  $\theta-\beta$ 
        if G.marks[theta, beta] != 0:
            continue
        # orient  $\theta \circ \rightarrow \beta$ : arrow at  $\beta$ , circle at  $\theta$ 
        if not (G.marks[theta, beta] == 1 and G.marks[beta, theta] == 0 and
                G.orient_with_homology(theta, beta, 1, 0) == 0):
            G.orient_with_homology(theta, beta, 1, 0)
            changed = True

# ----- R4 -----
# R4: If  $u = \langle \theta, \dots, \alpha, \beta, \gamma \rangle$  is a discriminating path between  $\theta$  and  $\gamma$ 
# and  $\beta \circ \rightarrow \ast \gamma$ , then:
# - if  $\beta \in \text{Sepset}(\theta, \gamma)$ , orient  $\beta \circ \rightarrow \ast \gamma$  as  $\beta \rightarrow \gamma$ 
# - else orient  $\langle \alpha, \beta, \gamma \rangle$  as  $\alpha \leftrightarrow \beta \leftrightarrow \gamma$ .
for beta in range(n):
    for theta in range(n):
        if theta == beta:
            continue
    for gamma in range(n):
        if gamma in (theta, beta):
            continue
        if not G.adj[beta, gamma]:
            continue
        #  $\beta \circ \rightarrow \ast \gamma$  : circle at  $\beta$  on edge  $(\beta, \gamma)$ 
        if G.marks[beta, gamma] != 0:
            continue
        # search discriminating paths  $\theta \dots \beta \gamma$ 
        for path in all_simple_paths_unbounded(theta, gamma):
            if beta not in path:
                continue
            if not is_discriminating_path(path, beta, theta, gamma):
                continue

```

```

# found discriminating path
sepset = G.get_sepset(theta, gamma)
if beta in sepset:
    #  $\beta \rightarrow \gamma$ 
    if not (G.marks[beta, gamma] == 1 and G.mark
            G.orient_with_homology(beta, gamma, 1, -
            changed = True
else:
    #  $\alpha \leftrightarrow \beta \leftrightarrow \gamma$ , where  $\alpha$  is predecessor of  $\beta$ 
    b_idx = path.index(beta)
    if b_idx == 0:
        continue
    alpha = path[b_idx - 1]
    #  $\alpha \leftrightarrow \beta$ 
    if not (G.marks[alpha, beta] == 1 and G.mark
            G.orient_with_homology(alpha, beta, 1, 1
            changed = True
    #  $\beta \leftrightarrow \gamma$ 
    if not (G.marks[beta, gamma] == 1 and G.mark
            G.orient_with_homology(beta, gamma, 1, 1
            changed = True
    break # only need one discriminating path

# ----- R5 -----
# If  $\alpha - \beta$  (undirected) and exists  $\gamma \rightarrow \alpha$  with  $\gamma$  not adjacent to  $\beta$ 
# orient  $\beta \rightarrow \alpha$  (arrowhead at  $\alpha$ ).
for alpha in range(n):
    for beta in range(n):
        if alpha == beta:
            continue
        #  $\alpha - \beta$  (both tails)
        if not (G.adj[alpha, beta] and
                G.marks[alpha, beta] == -1 and
                G.marks[beta, alpha] == -1):
            continue
        for gamma in range(n):
            if gamma in (alpha, beta):
                continue
            #  $\gamma \rightarrow \alpha$  (arrowhead at  $\alpha$ )
            if not (G.adj[gamma, alpha] and G.marks[gamma, alpha] == 1):
                continue
            #  $\gamma$  not adjacent to  $\beta$ 
            if G.adj[gamma, beta]:
                continue
            # orient  $\beta \rightarrow \alpha$  : tail at  $\beta$ , arrow at  $\alpha$ 
            if not (G.marks[beta, alpha] == 1 and G.marks[alpha, beta] == -1):
                G.orient_with_homology(beta, alpha, 1, -1)
                changed = True
            break

# ----- R6 -----
# If  $\alpha - \beta$  and there exists a discriminating path for  $\alpha$  between
# some  $\gamma$  and  $\beta$ , orient  $\beta \rightarrow \alpha$ .
for alpha in range(n):
    for beta in range(n):
        if alpha == beta:
            continue

```

```

#  $\alpha - \beta$ 
if not (G.adj[alpha, beta] and
        G.marks[alpha, beta] == -1 and
        G.marks[beta, alpha] == -1):
    continue

# search discriminating paths  $\langle \gamma, \dots, \alpha, \beta \rangle$ 
for gamma in range(n):
    if gamma in (alpha, beta):
        continue

    for path in all_simple_paths_unbounded(gamma, beta):
        # path must end with  $\dots, \alpha, \beta$ 
        if len(path) < 3:
            continue
        if path[-2] != alpha:
            continue

        if is_discriminating_path(path, alpha, gamma, beta):
            # orient  $\beta \rightarrow \alpha$ 
            if not (G.marks[beta, alpha] == 1 and G.marks[alpha, beta] == -1):
                G.orient_with_homology(beta, alpha, 1, -1)
                changed = True
            break
        if changed:
            break

# ----- R7 -----
# If  $\alpha - \beta$  cannot be oriented by R5 or R6 but remains undirected
# orient  $\alpha \leftrightarrow \beta$  (bidirected).
for alpha in range(n):
    for beta in range(n):
        if alpha == beta:
            continue
        # still undirected?
        if not (G.adj[alpha, beta] and
                G.marks[alpha, beta] == -1 and
                G.marks[beta, alpha] == -1):
            continue
        # convert to bidirected  $\alpha \leftrightarrow \beta$ 
        G.orient_with_homology(alpha, beta, 1, 1)
        changed = True

# ----- R8 -----
# R8: If  $\alpha \rightarrow \beta \rightarrow \gamma$  or  $\alpha \circ \beta \rightarrow \gamma$ , and  $\alpha \circ \rightarrow \gamma$ ,
# orient  $\alpha \circ \rightarrow \gamma$  as  $\alpha \rightarrow \gamma$ .
for alpha in range(n):
    for beta in G.neighbors(alpha):
        for gamma in G.neighbors(beta):
            if gamma == alpha:
                continue
            #  $\beta \rightarrow \gamma$ 
            if not is_arrow_into(gamma, beta):
                continue
            # either  $\alpha \rightarrow \beta$  or  $\alpha \circ \beta$ 
            cond1 = is_arrow_into(beta, alpha) #  $\alpha \rightarrow \beta$ 

```

```

cond2 = (G.adj[alpha, beta] and
         G.marks[alpha, beta] == 0 and # circle at
         G.marks[beta, alpha] == -1) # tail at  $\alpha$ 
if not (cond1 or cond2):
    continue
#  $\alpha \circ \rightarrow \gamma$  : circle at  $\alpha$ , arrow at  $\gamma$ 
if not (G.adj[alpha, gamma] and
        G.marks[alpha, gamma] == 1 and
        G.marks[gamma, alpha] == 0):
    continue
# orient  $\alpha \rightarrow \gamma$ : tail at  $\alpha$ , arrow at  $\gamma$ 
if not (G.marks[alpha, gamma] == 1 and G.marks[gamma,
        G.orient_with_homology(alpha, gamma, 1, -1)
        changed = True

# ----- helpers for R9-R10: uncovered p.d. paths -----

def is_uncovered(path):
    # every consecutive triple unshielded:  $V_{i-1}$  and  $V_{i+1}$  nonadjacent
    if len(path) < 3:
        return True
    for i in range(1, len(path) - 1):
        if G.adj[path[i - 1], path[i + 1]]:
            return False
    return True

def is_pd_edge(u, v):
    # edge  $u \rightarrow v$  is potentially directed from  $u$  to  $v$ 
    # if there's no arrowhead into  $u$  along  $u \rightarrow v$ 
    return G.adj[u, v] and (G.marks[v, u] != 1)

def uncovered_pd_paths(start, end):
    stack = [(start, [start])]
    while stack:
        (v, path) = stack.pop()
        for w in G.neighbors(v):
            if w in path:
                continue
            if not is_pd_edge(v, w):
                continue
            new_path = path + [w]
            if not is_uncovered(new_path):
                continue
            if w == end:
                yield new_path
            else:
                stack.append((w, new_path))

# ----- R9 -----
# R9: If  $\alpha \circ \rightarrow \gamma$ , and there is an uncovered p.d. path
#  $p = \langle \alpha, \beta, \theta, \dots, \gamma \rangle$  from  $\alpha$  to  $\gamma$  such that  $\beta$  and  $\gamma$ 
# are not adjacent, then orient  $\alpha \circ \rightarrow \gamma$  as  $\alpha \rightarrow \gamma$ .
for alpha in range(n):
    for gamma in range(n):
        if gamma == alpha:
            continue

```

```

#  $\alpha \circ \rightarrow \gamma$ 
if not (G.adj[alpha, gamma] and
        G.marks[alpha, gamma] == 1 and
        G.marks[gamma, alpha] == 0):
    continue
for path in uncovered_pd_paths(alpha, gamma):
    if len(path) < 3:
        continue
    beta = path[1]
    if not non_adjacent(beta, gamma):
        continue
    # orient  $\alpha \rightarrow \gamma$ 
    if not (G.marks[alpha, gamma] == 1 and G.marks[gamma,
        G.orient_with_homology(alpha, gamma, 1, -1)
        changed = True
    break # one path is enough

# ----- R10 -----
# R10: Suppose  $\alpha \circ \rightarrow \gamma$ ,  $\beta \rightarrow \gamma \leftarrow \theta$ ,
#       p1 is uncovered p.d. path  $\alpha \dots \beta$ ,
#       p2 is uncovered p.d. path  $\alpha \dots \theta$ ,
#       let  $\mu$  be neighbor of  $\alpha$  on p1,  $\omega$  neighbor of  $\alpha$  on p2,
#       if  $\mu \neq \omega$  and  $\mu, \omega$  nonadjacent, orient  $\alpha \circ \rightarrow \gamma$  as  $\alpha \rightarrow \gamma$ .
for alpha in range(n):
    for gamma in range(n):
        if gamma == alpha:
            continue
        #  $\alpha \circ \rightarrow \gamma$ 
        if not (G.adj[alpha, gamma] and
                G.marks[alpha, gamma] == 1 and
                G.marks[gamma, alpha] == 0):
            continue
        # nodes  $\beta, \theta$  with  $\beta \rightarrow \gamma \leftarrow \theta$ 
        parents = [v for v in range(n) if is_arrow_into(gamma, v)]
        for beta in parents:
            for theta in parents:
                if theta == beta or theta == alpha or beta == al
                    continue
                # p1: uncovered p.d. from  $\alpha$  to  $\beta$ 
                p1_list = list(uncovered_pd_paths(alpha, beta))
                if not p1_list:
                    continue
                # p2: uncovered p.d. from  $\alpha$  to  $\theta$ 
                p2_list = list(uncovered_pd_paths(alpha, theta))
                if not p2_list:
                    continue
                # take first such paths
                p1 = p1_list[0]
                p2 = p2_list[0]
                if len(p1) < 2 or len(p2) < 2:
                    continue
                mu = p1[1]
                omega = p2[1]
                if mu == omega:
                    continue
                if not non_adjacent(mu, omega):

```

```

        continue
        # orient  $\alpha \rightarrow \gamma$ 
        if not (G.marks[alpha, gamma] == 1 and G.marks[gamma, alpha] == 1):
            G.orient_with_homology(alpha, gamma, 1, -1)
            changed = True
        break #  $\beta, \theta$  found that trigger R10
    if changed:
        break
    # end loops over beta, theta
# end while changed

# --- main entry point ---

def fit(self, X, var_names=None):
    """
    X: numpy array (T, k)
    var_names: list of length k
    """
    X = np.asarray(X)
    T, k = X.shape
    if var_names is None:
        var_names = [f"X{i}" for i in range(k)]
    self.var_names_ = var_names

    # build lagged matrix
    Z, lagged_names = self._build_lagged_matrix(X, var_names)
    self.Z_ = Z

    # init dynamic PAG
    G = DynamicPAG(lagged_names, self.max_lag)

    assert G.n_nodes == Z.shape[1], (
        f"ERROR: Graph has {G.n_nodes} nodes but Z has {Z.shape[1]} columns."
        f"Check R5–R7 indentation and DynamicPAG construction.")

    # Algorithm 3.1 steps
    # 1–2: done via initialization
    # 3–8: skeleton
    self._skeleton_phase(G, Z)

    # 9: time orientation
    self._time_orientation(G)

    # 10: v-structures
    self._orient_v_structures(G)

    # 11: pds_s deletion
    self._pds_deletion_phase(G, Z)

    # 12: reset o-o and repeat 9–10
    G.reset_all_to_oo()
    self._time_orientation(G)
    self._orient_v_structures(G)

    # 13: R1–R10

```



```
self._apply_R_rules(G)
```

```
self.graph_ = G
return self
```

```
In [23]: # -----
# 5. ICF/BIC scoring via R ggm
# -----

import rpy2.robj as ro
from rpy2.robj import numpy2ri
numpy2ri.activate()

# load ggm + helper once
ro.r('library(ggm)')
ro.r("""
icf_bic <- function(S, amat, n) {
  A <- AG(amat, showmat=FALSE)
  fit <- fitAncestralGraph(S, A, n.obs=n)
  return(list(
    loglik = fit$loglik,
    df      = fit$df,
    bic     = -2*fit$loglik + fit$df * log(n)
  ))
}
""")
icf_bic_R = ro.r["icf_bic"]

def pag_to_mag(self, G):
    """
    Convert a PAG into a single valid MAG consistent with:
    - all invariant arrowheads
    - all invariant tails
    - ancestral graph constraints
    - minimal additional orientation

    Returns an adjacency matrix with codes:
    0 = no edge
    1 = i -> j    (directed)
    2 = i <-> j   (bidirected / latent confounding)
    3 = i - j     (undirected adjacency)
    """

    p = G.num_nodes
    amat = np.zeros((p, p), dtype=int)

    def has_arrowhead(i, j):
        return G.marks[i, j] == 1    # arrowhead at j

    def has_tail(i, j):
        return G.marks[i, j] == -1   # tail at j (i *- j)

    for i in range(p):
        for j in range(i + 1, p):
            if not G.adj[i, j]:
```

```

        continue

    # Case 1: Fully oriented edges
    if has_tail(i, j) and has_arrowhead(j, i):
        # i *-> j
        amat[i, j] = 1 # i->j
        continue
    if has_tail(j, i) and has_arrowhead(i, j):
        # j *-> i
        amat[j, i] = 1
        continue

    # Case 2: Invariant arrowheads (one-ended orientation)
    if has_arrowhead(i, j) and not has_arrowhead(j, i):
        # i *--> j means j is not ancestor of i, so orient i <- j
        amat[j, i] = 1
        continue
    if has_arrowhead(j, i) and not has_arrowhead(i, j):
        amat[i, j] = 1
        continue

    # Case 3: Bidirected (latent confounding)
    # PAG: o-> or <-o or o-o could hide latent confounding.
    # If both endpoints uncertain (circles), assign <-> as safe.
    if G.marks[i, j] == 0 and G.marks[j, i] == 0:
        amat[i, j] = amat[j, i] = 2 # i <-> j
        continue

    # Case 4: Undirected edge ambiguous endpoints (tail-circle)
    # Use undirected as minimal encoding.
    amat[i, j] = amat[j, i] = 3

return amat

def icf_bic_score(Z, G: DynamicPAG):
    """
    Compute ICF/BIC for one SVAR-FCI PAG.
    Z: lagged data (n, p_nodes) (same Z used in SVAR_FCI)
    G: DynamicPAG
    """
    n = Z.shape[0]
    S = np.cov(Z, rowvar=False)
    amat = pag_to_mag(G)
    res = icf_bic_R(S, amat, n)
    return {
        "loglik": float(res[0][0]),
        "df": float(res[1][0]),
        "bic": float(res[2][0]),
    }

```

```

-----
RuntimeError                                     Traceback (most recent call last)
Cell In[23], line 5
      1 # -----
      2 # 5. ICF/BIC scoring via R ggm
      3 # -----
      4
----> 5 import rpy2.robj as ro
      6 from rpy2.robj import numpy2ri
      7 numpy2ri.activate()

File /opt/anaconda3/lib/python3.13/site-packages/rpy2/robj/__init__.py:2
      19 import rpy2.rinterface.lib.openrlib
      20 import rpy2.rlike.container as rlc
----> 22 from rpy2.robj.robj import RObjectMixin, RObject
      23 import rpy2.robj.functions
      24 from rpy2.robj.environments import (Environment,
      25                                     local_context)

File /opt/anaconda3/lib/python3.13/site-packages/rpy2/robj/robj.py:11
      7 import rpy2.rinterface.lib.callbacks
      9 from rpy2.robj import conversion
----> 11 rpy2.rinterface.intr()
      14 def _add_warn_reticulate_hook():
      15     msg = """
      16     WARNING: The R package "reticulate" only fixed recently
      17     an issue that caused a segfault when used with rpy2:
      18     (...)
      19     the fix.
      20     """
      21

File /opt/anaconda3/lib/python3.13/site-packages/rpy2/rinterface/__init__.py:1130, in intr(interactive, _want_setcallbacks, _c_stack_limit)
      1127     logger.info('R is already initialized. No need to initialize.')
      1128     return None
-> 1130 _setrenvvars(_ENVVAR_ACTION_MAP)
      1131 if embedded.is_r_externally_initialized():
      1132     embedded._setinitialized()

File /opt/anaconda3/lib/python3.13/site-packages/rpy2/rinterface/__init__.py:1222, in _setrenvvars(action_map)
      1220 def _setrenvvars(action_map: typing.Dict[str, _ENVVAR_ACTION]):
      1221     new_envvars = {}
-> 1222     for k, v in _getrenvvars():
      1223         if k in os.environ:
      1224             action = action_map[k]

File /opt/anaconda3/lib/python3.13/site-packages/rpy2/rinterface/__init__.py:1175, in _getrenvvars(baselinevars, r_home)
      1173     r_home = openrlib.R_HOME
      1174     if r_home is None:
-> 1175         raise RuntimeError('Unable to determine R_HOME.')
      1177 # Use a temporary file to write the environment variables. Windows
      1178 # has a file locking system that requires a slightly more complicate

```

```
d
1179 # implementation than it would otherwise be on other OSes.
1180 temp_fh = tempfile.NamedTemporaryFile(mode='w', delete=False, suffix
='.csv')

RuntimeError: Unable to determine R_HOME.
```

```
In [ ]: # -----
# 6. Data-driven selection of alpha and p (Appendix)
# -----

def select_alpha(
    X,
    var_names,
    p,
    alpha_grid=np.arange(0.01, 0.41, 0.01),
    verbose=False,
):
    """
    For fixed maximum lag p, choose alpha maximizing BIC(P_hat_alpha).
    """
    best_alpha = None
    best_bic = np.inf
    best_model = None
    best_score = None

    for alpha in alpha_grid:
        if verbose:
            print(f"alpha={alpha:.3f}")
        model = SVAR_FCI(alpha=alpha, max_lag=p, verbose=False)
        model.fit(X, var_names=var_names)
        score = icf_bic_score(model.Z_, model.graph_)
        if verbose:
            print(" BIC:", score["bic"])
        if score["bic"] < best_bic:
            best_bic = score["bic"]
            best_alpha = alpha
            best_model = model
            best_score = score

    return best_model, best_alpha, best_score

def select_p(
    X,
    var_names,
    alpha,
    p_grid,
    verbose=False,
):
    """
    For fixed alpha, choose p maximizing BIC(P_hat_p).
    """
    best_p = None
    best_bic = np.inf
    best_model = None
```

```

best_score = None

for p in p_grid:
    if verbose:
        print(f"p={p}")
    model = SVAR_FCI(alpha=alpha, max_lag=p, verbose=False)
    model.fit(X, var_names=var_names)
    score = icf_bic_score(model.Z_, model.graph_)
    if verbose:
        print("  BIC:", score["bic"])
    if score["bic"] < best_bic:
        best_bic = score["bic"]
        best_p = p
        best_model = model
        best_score = score

return best_model, best_p, best_score

def select_alpha_and_p(
    X,
    var_names,
    alpha_grid=np.arange(0.01, 0.41, 0.01),
    p_grid=range(1, 5),
    verbose=False,
):
    """
    Joint search over alpha and p as described in the appendix.
    """
    best_alpha = None
    best_p = None
    best_bic = np.inf
    best_model = None
    best_score = None

    for alpha in alpha_grid:
        for p in p_grid:
            if verbose:
                print(f"alpha={alpha:.3f}, p={p}")
            model = SVAR_FCI(alpha=alpha, max_lag=p, verbose=False)
            model.fit(X, var_names=var_names)
            score = icf_bic_score(model.Z_, model.graph_)
            if verbose:
                print("  BIC:", score["bic"])
            if score["bic"] < best_bic:
                best_bic = score["bic"]
                best_alpha = alpha
                best_p = p
                best_model = model
                best_score = score

    return best_model, best_alpha, best_p, best_score

```

```
In [ ]: import numpy as np
```

```
def simulate_svar_data(T=1000, seed=0):
```

```

"""
Simulate 3-var VAR(2) with known causal DAG:
    X_{t-1} → Y_t
    Y_{t-1} → Z_t
    Z_{t-1} → X_t
plus latent confounding: X_t ↔ Z_t

Returns:
    X : (T,3) matrix
    var_names : ["X","Y","Z"]
"""
rng = np.random.default_rng(seed)

A1 = np.array([
    [0.0, 0.0, 0.4], # Z_{t-1} → X_t
    [0.3, 0.0, 0.0], # X_{t-1} → Y_t
    [0.0, 0.2, 0.0]  # Y_{t-1} → Z_t
])
A2 = np.zeros((3,3)) # no second lag effect, but keeps structure simple

# latent confounding: U influences X_t and Z_t
T = T + 5
X = np.zeros((T,3))
for t in range(2, T):
    eta = rng.normal(size=3)
    latent = rng.normal()

    X[t] = (
        A1 @ X[t-1]
        + A2 @ X[t-2]
        + eta
        + np.array([latent, 0, latent]) * 0.5
    )

return X[5:], ["X","Y","Z"]

```

```

In [ ]: def test_svar_fci_once():
    X, names = simulate_svar_data(T=1200, seed=42)
    model = SVAR_FCI(alpha=0.05, max_lag=2, verbose=True)
    model.fit(X, var_names=names)

    G = model.graph_
    print("\n=== Learned PAG marks ===")
    print(G.marks)
    print("\n=== Learned adjacency ===")
    print(G.adj.astype(int))

    # Test MAP→MAG conversion
    amat = pag_to_mag(model, G)
    print("\n=== Converted MAG adjacency ===")
    print(amat)

    return model, amat

model, amat = test_svar_fci_once()

```

In [ ]: