```
<img style="margin-left: 5px;" src=head.png width=150 align=right>
       The next generation of AI-literature software engineers need to know how to mix-and-match the services within AI tools. To that end, I've been refactoring the work of my AI graduate students (3 dozen over 20 years) into a tool kit small enough to build in a semester, that can be refactored many ways. So my standard "intro to AI" project is to get students to rebuild the following code, from scratch, in any language they like (except LUA) in six weekly homeworks. <br/>
'br clear=all>
        Standard supervised learners assume that all examples have labels.
       When this is not true, then we need tools to incrementally (a) summarize what has been seen so far; (b) find and focus on the most interesting part of that summary, (c) collect more data in that region, then (d) repeat.
       <a href="div.png"><img align=right width=225 src="div.png"></a>
To make that search manageable, it is useful to exploit a
manifold assumption; i.e.
higher-dimensional data can be approximated in a lower dimensional
manifold without loss of signal [Ch05, Le05].
Manifolds lead to _continuity_
effects; i.e. if there are fewer dimensions, then there are more
similarities between examples.
Continuity simplifies _clustering_
(and any subsequent reasoning). More similarities means easier
clustering. And after clustering, reasoning just means reason about
a handful of examples (maybe even just one) from each cluster.
       To exploit the manifold assumption, this code uses Aha's distance measure [Aha91] (that can handle numbers and symbols) to recursively divide data based on two distant points (these are found in linear time using the Fastmap heuristic [Fa95]). To avoid spurious outliers, this code use the 90% furthest points. To avoid long runtimes, uses a subset of the data to learn where to divide data (then all the data gets pushed down first halves).
 -- **ASSIGNMENTS**
 Discretize the numeric ranges (\*) at each level of the recursion, then divide the data according what range best selects for one half, or the o
        at the data at this level of recursion.

- **Multi-objective optimization:** This code
can apply Zitzler's multi-objective rankining predicate [Zit04] to prune the
          st
half of the data, then recurs on the rest [Ch18]. Assuming a large over-gener
 action — of the initial population (to say, 10,000, examples), this can be just as eff
       as genetic optimization [Ch18], but runs much faster.
- **Semi-supervised learning**: these applications require only the _2.log(N)
        labels at of furthest points seen at each level of recursion.
 14 == a little LUA learner laboratory.
(c) 2022, Tim Menzies, BSD 2-clause license.
       -p
-seed
 KEY: N=fileName F=float P=posint S=string
NOTES: This code uses Aha's distance measure [Aha91] (that can handle numbers and symbols) to recursively divide data based on two distant points (these two are found in linear time using the Fastmap
 heuristic (Fa951).
To avoid spurious outliers, this code use the 90% furthest points.
To avoid long runtimes, uses a subset of the data to learn where to divide data ({\it then} all the data gets pushed down first halves).
To support explanation, optionally, at each level of recursion, this code reports what ranges can best distinguish sibling clusters C1, C2. The discretizer is inspired by the ChiMerge algorithm: numerics are divided into, say, 16 bins. Then, while we can find adjacent bins, merge them then look for other merges.
-- Cache current globals, use at end to find rogue variables local b4={}; for k,_ in pairs(_ENV) do b4[k]=k end
  -- Defined local names
local any, asserts, big, cli, csv, fails, firsts, fmt, goalp, ignorep, klassp local lessp, map, main, many, max, merge, min, morep, new, nump, o, oo, per, pop, push local r, rows, slots, sort, sum, thing, things, unpack
 -- Classes have UPPER CASE names.
local CLUSTER, COLS, EGS, NUM, ROWS = {},{},{},{},{},{}
local SKIP, SOME, SPAN, SYM = {},{},{},{}
       ## Settings
Parse the help text for flags and defaults (e.g. -keep, 512).
Check for updates on those details from command line
(and and there,
some shortcuts are available;
e.g. _-k N_ ⇒ 'keep=N';
and _-booleanFlag_ ⇒ 'booleanFlag=not default').
-- this code reads csv files where the words on linel define column types.

function ignorep(x) return x:find".5" end -- columns to ignore
function lessp(x) return x:find".5" end -- symbolic goals to achieve
function morep(x) return x:find".5" end -- number goals to minimize
function nump(x) return x:find".5" end -- numeric goals to maximize
function nump(x) return x:find".5" end -- numeric goals to maximize
function goalp(x) return morep(x) or lessp(x) or klassp(x) end
 -- strings
fmt = string.format
```

```
-- maths
-- maths
-- math.huge
-- math.max
```

```
function new(k,t) k.__index=k; k.__tostring=o; return setmetatable(t,k) end
-- COLS: turns list of column names into NUMs, SYMs, or SKIPs function COLS.new(k,row, i) i= new(k, (all={{1}},x={{1}},y={{1}},names=row{{1}}) for at,txt in ipairs(row) do push(i.all, i:col(at,txt)) end return i end
 function COLS.add(i,t)
     for _,col in pairs(i.all) do col:add(t[col.at]) end return t end
function COLS.col(i,at,txt, col)
if ignorep(txt) then return SKIP:new(at,txt) end
col = (nump(txt) and NUM or SYM):new(at,txt)
push(goalp(txt) and i.y or i.x, col)
if klassp(txt) then i.klass = col end
return col end
 -- NUM: summarizes a stream of numbers
function NUM.new(k,n,s)
return new(k,fn=0,at=n or 0,txt=s or"",has=SOME:new(),ok=false,
w=lessp(s or "") and -1 or 1, lo=big, hi=-big}) end
function NUM.add(i,x)
  if x ~= "?" then
  i.n = i.n + 1
  if i.has:add(x) then i.ok=false end
  i.lo,i.hi = min(x,i.lo), max(x,i.hi); end end
function NUM.mid(i) return per(i:sorted(), .5) end
function NUM.norm(i,x)
  return math.abs(i.hi-i.lo)<1E-9 and 0 or (x-i.lo)/(i.hi - i.lo) end</pre>
 function NUM.sorted(i)
  if i.ok==false then table.sort(i.has.all); i.ok=true end
  return i.has.all end
 -- ROWS: manages 'rows', summarized in 'cols' (columns).

function ROWS.new(k,inits, i)

i = new(k,{rows={},cols=nil})

if type(inits)=="string" then for t in csv(inits) do i:add(t) end end
if type(inits)=="table" then for t in inits do i:add(t) end end
return i end
function ROWS.add(i,t)
  if i.cols then push(i.rows,i.cols:add(t)) else i.cols=COLS:new(t) end end
 function ROWS.clone(i, j) j= ROWS:new(); j:add(i.cols.names);return j end
function ROWS.dist(i,row1,row2,     d,fun)
   function fun(col) return col:dist(row1[col.at], row2[col.at])^the.p end
   return (sum(i.cols.x, fun)/ #i.cols.x)^(1/the.p) end
function ROWS.furthest(i,row1,rows, fun)
  function fun(row2) return (i:dist(row1,row2), row2) end
  return unpack(per(sort(map(rows,fun),firsts), the.furthest)) end
function ROWS.half(i, top)
local some, top,c,x,y,tmp,mid,lefts,rights,_
some= many(i.rows, the.keep)
top = top or i
    _,x = top:furthest(any(some), some)
    c,y = top:furthest(x, some)
tmp = sort(map(i.rows,function(r) return top:fastmap(r,x,y,c) end),firsts)
mid = #i.rows//2
lefts, rights = i:clone(), i:clone()
for at,row in pairs(tmp) do (at <=mid and lefts or rights):add(row[2]) end
return lefts,rights,x,y,c, tmp[mid] end</pre>
 function ROWS.mid(i,cols)
  return map(cols or i.cols.all, function(col) return col:mid() end) end
 -- SKIP: summarizes things we want to ignore (so does nothing) function SKIP.new(k,n,s) return new(k,(n=0,at=at or 0,txt=s or""}) end function SKIP.add(i,x) return x end function SKIP.mid(i) return "?" end
 -- SOME: keeps a random sample on the arriving data function SOME.new(k, keep) return new(k, {n=0,all={}}, keep=keep or the.keep}) end function SOME.add(i,x)
                    #i.all < i.keep then push(i.all,x) ; return i.all
r() < i.keep/i.n then i.all[r(#i.all)]=x; return i.all end end</pre>
 -- SYM: summarizes a stream of symbols
function SYM.new(k,n,s)
  return new(k, {n=0,at=n or 0,txt=s or"",has={},most=0}) end
function SYM.add(i,x,inc)
   if x -= "?" then
   inc = inc or 1
   i.n = i.n + inc
   i.has[x] = inc + (i.has[x] or 0)
   if i.has[x] > i.most then i.most,i.mode=i.has[x],x end end end
 function SYM.dist(i,x,y) return(x=="?" and y=="?" and 1) or(x==y and 0 or 1) end
function SYM.div(i, return i.mode end
function SYM.div(i, p)
return sum(i.has,function(k) p=-i.has[k]/i.n;return -p*math.log(p,2) end) end
     inction SYM.merge(i,j, k)
k = SYM:new(i.at,i.txt)
for x,n in pairs(i.has) do k:add(x,n) end
for x,n in pairs(j.has) do k:add(x,n) end
ei, ej, ejk= i:div(), j:div(), k:div()
if i.n==0 or j.n==0 or .99*ek <= (i.n*ei + j.n*ej)/k.n then
return k end end</pre>
```

```
--- CLUSTER: recursively divides data by clustering towards two distant points
function CLUSTER.new(k, sample, top)
local i, enough, left, right
     local i,enough,left,right
top = top or sample)
i = new(K, {here=sample})
enough = (#top.rows)^the.enough
if #sample.rows >= 2*enough then
left, right, i.x, i.y, i.c, i.mid = sample:half(top)
if #left.rows < #sample.rows then
i.left = CLUSTER:new(left, top)
i.right= CLUSTER:new(right, top) end end
return i end</pre>
function CLUSTER.show(i,pre, here)
     inction CLUSTEK.show(i,pre, nere)
pre = pre or ""
here=""
if not i.left and not i.right then here= o(i.here:mid(i.here.cols.y)) end
print(fmt("%6:%-30s%", #i.here.rows, pre, here))
for _kid in pairs(i.left, i.right) do
    if kid then kid:show(pre .. "[.") end end end
 -- SPAN: keeps a random sample on the arriving data function SPAN.new(k, col, lo, hi, has) return new(k, col-col,lo=lo,hi=hi or lo,has=has or SYM:new()}) end
 function SPAN.add(i,x,y,n) i.lo,i.hi=min(x,i.lo),max(x,i.hi); i.has:add(y,n) end
function SPAN.merge(i,j)
local has = i.has:merge(j.has)
if now then return SPAN:new(i.col, i.lo, j.hi, has) end end
 function SPAN.select(i,row, x)
      return (x=="?") or (i.lo==i.hi and x==i.lo) or (i.lo <= x and x < i.hi) end</pre>
 -- EXPLAIN

function EXPLAIN(k, sample, top)
    i.here = sample
    top = top or sample
    enough = (#top.rows)^the.enough
    if #top.rows >= 2*enough then
    left,right = sample:half(top)
    spans == {}
    for n,col in pairs(i.cols.x) do
        tmp = col:spans(j.cols.x[n])
        if #tmp>1 then for _,one in pairs(tmp) do push(spans,one) end end
        if #spans > 2 then
        XXXX?
 function SYM.spans(i, j)
local xys,all,one,last,xys,x,c n = {},{}
for x,n in pairs(i.has) do push(xys, {x, "this",n}) end
for x,n in pairs(j.has) do push(xys, {x, "that",n}) end
for _,tmp in ipairs(sort(xys,firsts)) do
    x,c,n = unpack(tmp)
    if x ~= last then
    last = x
        one = push(all, Span(i,x,x)) end
    one:add(x,y,n) end
return all end
```

```
fails=0
function asserts(test, msg)
print(test and "PASS: "or "FAIL: ",msg or "")
if not test then
fails=fails+1
if the.dump then assert(test,msg) end end end
.
          function EGS.nothing() return true end function EGS.the() oo(the) end function EGS.some(s,t) s=SOME:new(100)
410
                  s=SOME:new(100)
for i=1,100000 do s:add(i) end
for j,x in pairs(sort(s.all)) do
   --if (j % 10)==0 then print("") er
   --io.write(fmt("%6s",x)) end end
fmt("%6s",x) end end
                  r = ROWS:new(the.data)

r = ROWS:new(the.data)

s = r:clone()

for _,row in pairs(r.rows) do s:add(row) end
asserts(r.cols.x[1].lo==s.cols.x[1].lo, "clone.lo")
asserts(r.cols.x[1].hi==s.cols.x[1].hi, "clone.hi")
end
            function EGS.clone( r.s)
          function EGS.data( r)
  r = ROWS:new(the.data)
  asserts(r.cols.x[1].hi == 8, "data.columns") end
           function EGS.dist( r,rows,n)
r = ROWS:new(the.data)
                  r = ROWS:new(tne.uara,
rows = r.rows
n = NUM:new()
for _,row in pairs(rows) do n:add(r:dist(row, rows[1])) end
--oo(r.cols.x[2]:sorted()) end
          function EGS.many(     t)
t={}; for j=1,100 do push(t,j) end
--print(oo(many(t, 10))) end
o(many(t, 10)) end
          function EGS.far(    r,c,row1,row2)
    r = ROWS:new(the.data)
    row1 = r.rows[1]
    c,row2 = r:far(r.rows[1], r.rows) end
    --print(c,"\n",o(row1),"\n", o(row2)) end
           function EGS.half( r,c,row1,row2)
local lefts,rights,x,y,x
r = ROWS:new(the.data)
r:mid(r.cols.y)
                  lefts, rights, x, y, c = r:half()
lefts:mid(lefts.cols.y)
rights:mid(rights.cols.y)
asserts(true, "half") end
                         = ROWS:new(the.data)
                  --CLUSTER:new(r):show() end
CLUSTER:new(r) end
         for k,v in pairs(_ENV) do if not b4[k] then print("?",k,type(v)) end end
os.exit(fails)
                 return {CLUSTER=CLUSTER, COLS=COLS, NUM=NUM, ROWS=ROWS, SKIP=SKIP, SOME=SOME, SYM=SYM,the=the,oo=oo,o=o}
                     git rid of SOME for rows
nss = NUM | SYM | SKIP
COLS = all:[nss]+, x:[nss]*, y:[nss]*, klass;col?
ROWS = cols:COLS, rows:SOME
         -- ROWS = cols:COLS, rows:SOME
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