Subject: Re: Avoiding multiple FOR loops
Posted by Jeremy Bailin on Sat, 19 Mar 2011 04:35:43 GMT
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The problem, I think, is the regress statement. Everything else can be completely vectorized, but you're going to have to loop through all the combinations and run the regression on each. Still, this may save time:

1. Generate the list of combinations. E.g.

```
nper = indgen(nvar) + maxn - nvar
vars = lindgen(nper)
vars_ai = array_indices(vars, vars)  ; all permutations with duplicates
goodcombip = vars_ai[0:nvar-2,*] lt vars_ai[1:*,*]
goodcombi = where(total(goodcombip, 1, /int) eq nvar-1, ngoodcombi)
vars_ai = vars_ai[*, goodcombi]  ; all unique combinations
```

(note: I'm sure there's a better way to do this, but it's not going to be the limiting step so this is probably good enough)

2. Collect the data

```
x = data[vars_ai, *]
```

3. Loop through the regressions and store yfit.

```
ndata = (size(x, /dimen))[1]
yfits = fltarr(ndata, ngoodcombi]
for i=0l, ngoodcombi-1 do begin
  coef = regress(x[i*nvar:(i+1)*nvar-1, *], y, const=const, yfit=yfit)
  yfits[*,i] = yfit
endfor
```

4. Calculate rms.

```
rms = sqrt(total(( rebin(y,ndata,ngoodcombi,/sample) - yfits)^2, 1) / ndata)
```

Note: totally untested and totally untimed, so I don't know if this is any faster. My gut feeling is that having one for loop that has a minimal amount of code in it is going to be faster than 6 nested for loops that have a few more lines in the meat, but I don't know by how much and it certainly won't be nearly as fast as if regress were vectorized.

-Jeremy.

Subject: Re: Avoiding multiple FOR loops
Posted by Gray on Sat, 19 Mar 2011 11:06:43 GMT
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On Mar 19, 12:35 am, Jeremy Bailin <astroco...@gmail.com> wrote:
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> -Jeremy.

Any way you can use mpfit?

Subject: Re: Avoiding multiple FOR loops
Posted by fgg on Tue, 22 Mar 2011 01:51:43 GMT
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Thanks a lot, Jeremy. I'll test this code to see how faster it is.

Gray, I'm not really familiar with it. But is doesn't seem to include routines for variable selection. Does it?

Thanks!