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"""
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```
coadd(sc, D)
```

Co-add a sampled up-the-ramp datacube into a SIRSCore.

```
Parameters: sc::SIRSCore
```

```
    A SIRSCore struct
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```
    D::Array{Float64,3}
```

```
    An HxRG datacube
```

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"""
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```
function coadd!(sc::SIRSCore, D::Array{Float64,3})
```

```
# Compute residuals by fitting and subtracting a straight line.
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```
# A bit of testing shows that using MKL to do the matrix multiplications is
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```
# faster than using mapslices. In other contexts, I have found
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```
# that using MKL is faster than tensor operations. Therefore,
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```
# overwrite D so that it will work with MKL.
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```
D = permutedims(D, (3,1,2)) # This is expensive, on Racy ~14 seconds but
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```
                        # done only once per exposure
```

```
 $\Delta$  = D - reshape(sc.L_x_Linv * reshape(D, (sc.naxis3,sc.naxis1*sc.naxis2)),  
                  (sc.naxis3,sc.naxis1,sc.naxis2))
```

```
# Work in frames
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```
for z in 1:sc.naxis3
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```
    # Pick out one frame
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```
    frm =  $\Delta$ [z,:,:]
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```
    # Get a copy of the good pixel mask. Make a copy so
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```
    # as not to mess it up.
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```
    gdp_x = copy(sc.gdp_x)
```

```
    # Compute incomplete Fourier transforms of reference pixels.
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```
    # Note added 2/21/22. I tried parallelizing these two lines on Racy.
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```
    # There was no speed improvement. The overheads associated with starting
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    # the additional threads were larger than the improvement.
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```
    # From timing the code, these are actually very fast.
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```
     $\ell$  = inc_rfft(sc.SFT, reshape(frm[1:sc.rb,:], :))
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```
     $r$  = inc_rfft(sc.SFT, reshape(frm[end-sc.rb+1:end,:], :))
```

```
# Loop over outputs
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```
Threads.@threads for op in 1:sc.nout
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```
    # Get column range for this output
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c1 = (op-1) * sc.xsize + 1
c2 = c1 + sc.xsize - 1
crng = c1:c2

# Get data for this output.
# 1) We overwrite all elements, so it is faster to leave this
#     uninitialized.
# 2) Include NROH additional columns for the new row overhead
#     to eventually hold new row overhead
# 3) Flip axis 1 of even numbered columns
d = Array{Float64, 2}(undef, (sc.xsize+sc.nroh,sc.ysize))
if mod(op,2) != 0
    d[1:sc.xsize,:] = frm[crng,:] # Don't flip
else
    d[1:sc.xsize,:] = frm[crng,:][end:-1:1,:] # Flip
end

# Get the good pixel map for this output flipping
# axis 1 as necessary
_gdpx = BitArray{2}(Array{Bool,2}(undef, (sc.xsize,sc.ysize)))
if mod(op,2) != 0
    _gdpx = gdpx[crng,:]
else
    _gdpx = gdpx[crng,:][end:-1:1,:]
end

# Find and flag transients. This only makes sense
# in the regular pixels. Known bad pixels are already
# flagged. Trim thresholds are defined in SIRS.jl.
if (op==1) || (op==32)
    regpix_mask = sc.regpix_edge
else
    regpix_mask = sc.regpix_middle
end
regular_pixels = d[1:sc.xsize,:][regpix_mask]
good_pixels = _gdpx[regpix_mask]
sorted = sort(regular_pixels[good_pixels])
nrej = Int64(round(TRIM_HARD * length(sorted))) # Figure out how
                                                # many to trim
                                                # on either side

min_good = sorted[nrej]
max_good = sorted[end-nrej+1]
good_pixels[regular_pixels .<= min_good] .= 0
good_pixels[regular_pixels .>= max_good] .= 0

```

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_gdpx[regpix_mask] = good_pixels

# Output level quality control.
# Compute the mean and standard deviation of remaining good pixels.
# Use this to find any anomalous behavior in this output and frame.
# Also compute the mean since it will be useful later.
here = d[1:sc.xsize,HXRG_RB+1:end-HXRG_RB]
μ = mean(here[_gdpx[:,HXRG_RB+1:end-HXRG_RB]])
# This seems to cause problems...
# if ! ((GD_OP_MU_MIN .< μ) * (μ .< GD_OP_MU_MAX))
#     println("Skipping frame ", z, " output ", op, " μ = ", μ)
#     flush(stdout)
#     continue
# end
σ = std(here[_gdpx[:,HXRG_RB+1:end-HXRG_RB]])
if ! ((GD_OP_SIG_MIN .< σ) * (σ .< GD_OP_SIG_MAX))
    println("Skipping frame ", z, " output ", op, " σ = ", σ)
    flush(stdout)
    continue
end

# Get the (robust) mean reference row value.
# From looking at a lot of data, the rows selected here tend to be
# most indicative of the regular pixels in Roman H4RGs.
μ_refrows = mean(trim(reshape((d[1:sc.xsize,
                                sc.naxis2-2:sc.naxis2-1]), :),
                    prop=TRIM_HARD))

# Replace known bad pixels and transients with the mean
# of all good pixels in the same row.
for y in 1:sc.ysize
    # Only touch rows that contain bad pixels
    if !(0 ∈ _gdpx[:,y])
        continue
    end
    here = d[1:sc.xsize,y]
    here[_gdpx[:,y] .== 0] .= mean(here[_gdpx[:,y] .== 1])
    d[1:sc.xsize,y] = here
end

# Fill overhead columns by mirroring
roi = d[end-2sc.nroh+1:end-sc.nroh,:]
d[end-sc.nroh+1:end,:] = roi[end:-1:1,:]

```

```

# Go to vectors. View the reference pixel stream as an xy-plot.
# The x-axis is pixel index and the y-axis is signal in DN.
d_vec = reshape(d,:)

# Compute the FFT
n = rfft(d_vec)

# Keep just frequencies of interest. These are less than Nyquist
# on the row rate and within the same frequency interval of Nyquist.
n = cat(n[1:sc.naxis2÷2+1],n[length(n)-(sc.naxis2÷2-1):end],
        dims=1)

# Coadd sums for frequencies > 0 Hz
sc.N[:,op] .+= real.(n .* conj(n))
sc.L[:,op] .+= real.(l .* conj(l))
sc.R[:,op] .+= real.(r .* conj(r))
sc.X[:,op] .+= n .* conj(r)
sc.Y[:,op] .+= n .* conj(l)
sc.Z[:,op] .+= r .* conj(l)

# Coadd sums for f = 0 Hz
sc.R[op]    += μ - μ_refrows
sc.N[op]    += 1           # Serves as frame counter

end
end
end

```