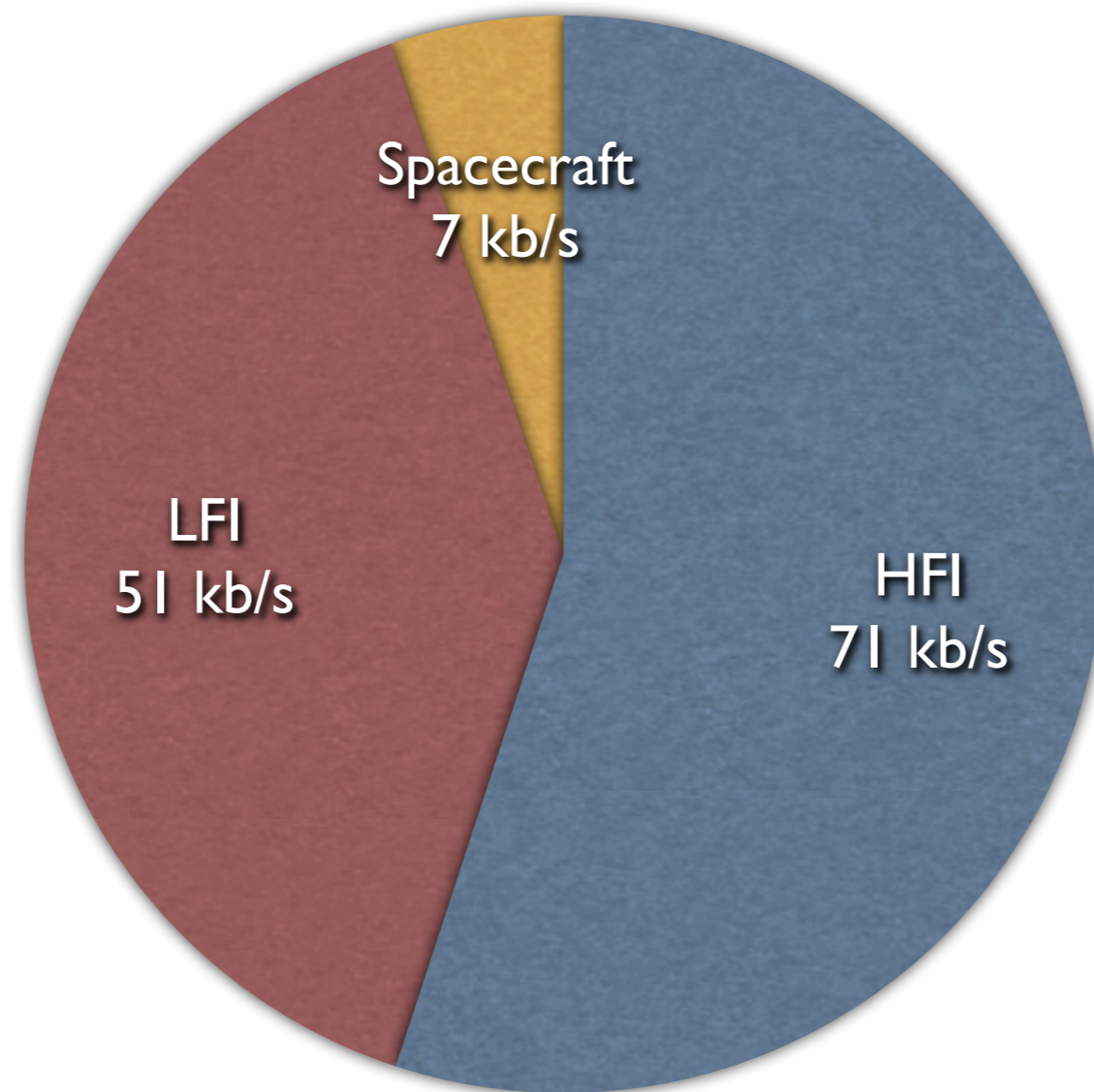


Beyond PKZip: Data compression for Planck/LFI

Maurizio Tomasi
Astrosiesta INAF, April, 7th 2011

Why do we need a compressor for LFI?



LFI data need ~2.4 times more than 51 kb/s!

Part I

Principles of data compression

Types of compressors

The two categories of algorithms I am going to discuss today are:

- Dictionary compressors;
- Statistical compressors.

Dictionary

compressors: example

When you find this...	...substitute it with this
as far as I know	AFAIK
by the way	BTW
id est	ie.
exempli gratia	eg.
post scriptum	ps.

Dictionary compressors

- Principle: substitute recurring patterns of symbols with shorter sequences
- Many compression programs belong to this family: PKZip, RAR, GZip, BZip2...
- This compression is very sensitive to the order of the symbols in the input!

Statistical compressors

- Compressor of this type work on the bit sequences which encode each symbol.
- They compress better when some *symbols* (not patterns!) occur more than others.
- Popular example: the JPEG file format.

An example

1304 1304 1301 1302 1301 1303 1304 1304 (16x8=128 bit)

An example

1304 1304 1301 1302 1301 1303 1304 1304 (16x8=128 bit)



11 11 00 01 00 10 11 11 (2x8=16 bit)

Symbol	Bit mask
1301	00
1302	01
1303	10
1304	11

An example

11 11 00 01 00 10 11 11 (16 bit)



00 10 110 10 111 00 (14 bit)

Symbol	Frequency	Bit mask
11	4	0
00	2	10
01	1	110
10	1	111

Information entropy

- To estimate the performance of a statistical compressor, Shannon defined the so-called *information entropy*.
- If the i -th symbol occurs with frequency $p(i)$, then the information entropy of the signal is

$$H = - \sum_i p(i) \log_2 p(i)$$

(in our previous example, $H = 7/4$)

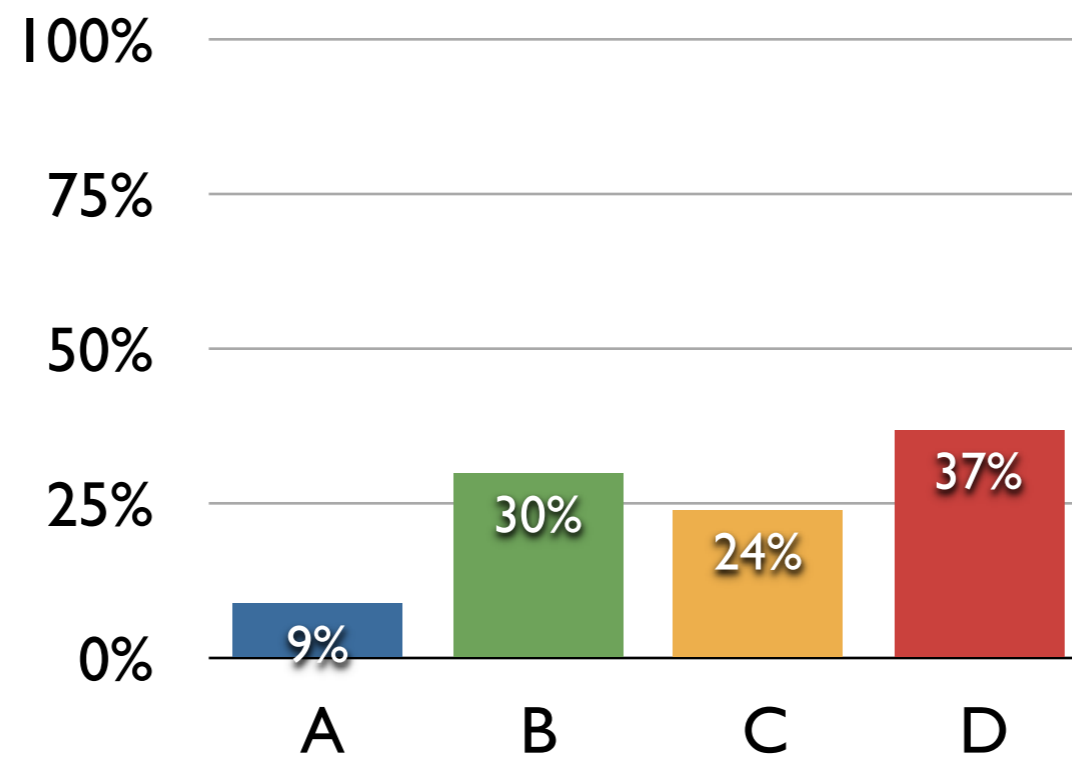
Maximum achievable compression ratio

The maximum compression ratio achievable by any statistical compressor can be calculated analytically:

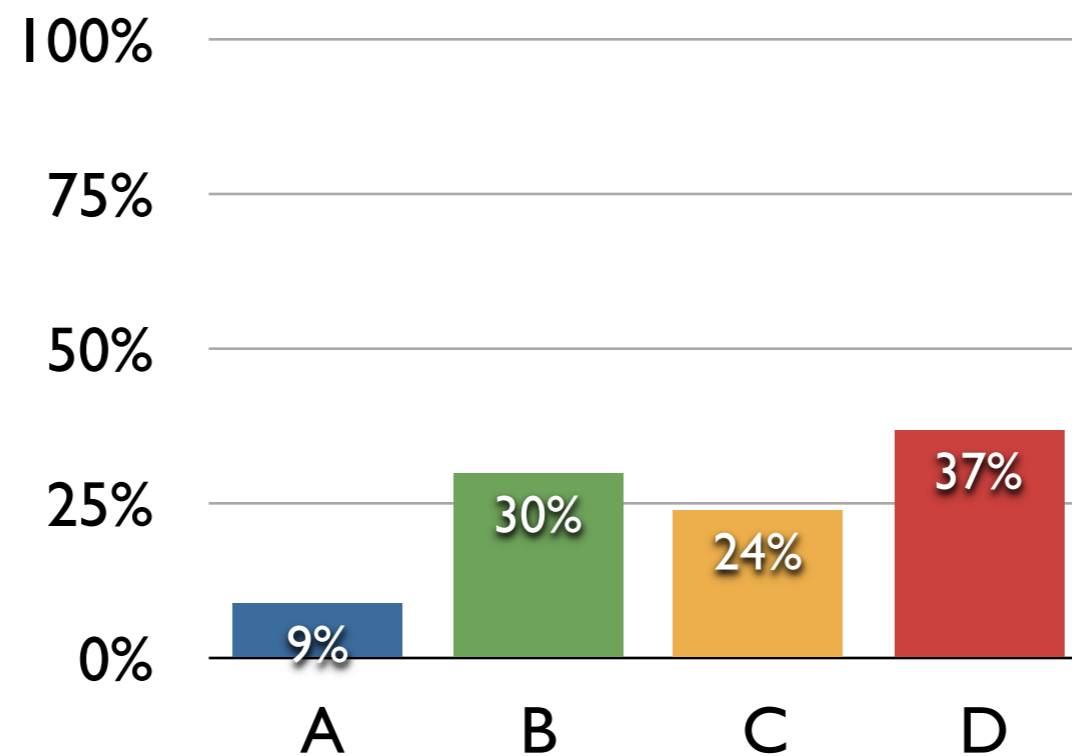
$$C_r^{\max} = \frac{n_{\text{bits}}}{H} = \frac{\log_2 N_{\text{symb}}}{H}$$

In our previous example we reached this limit ($16/14 \sim 1.142$). We can be proud!

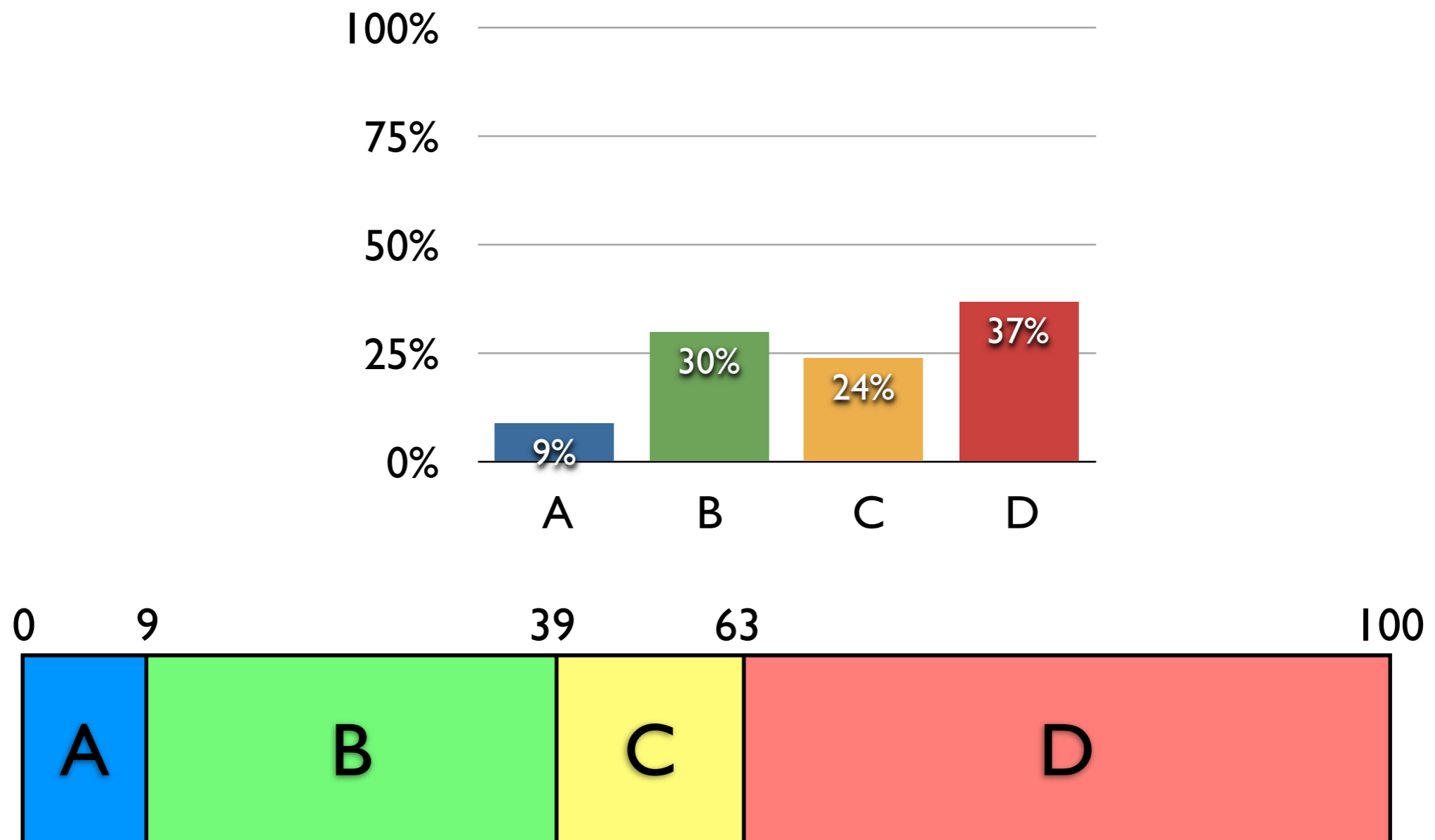
Arithmetic compression



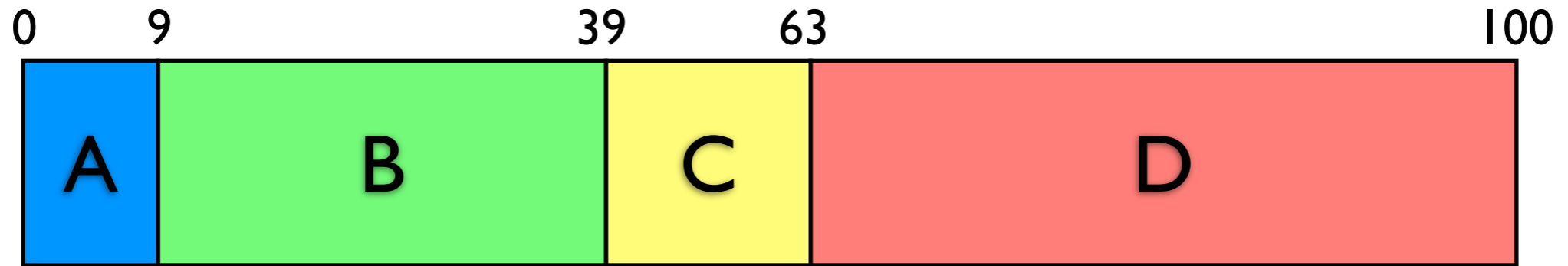
Arithmetic compression



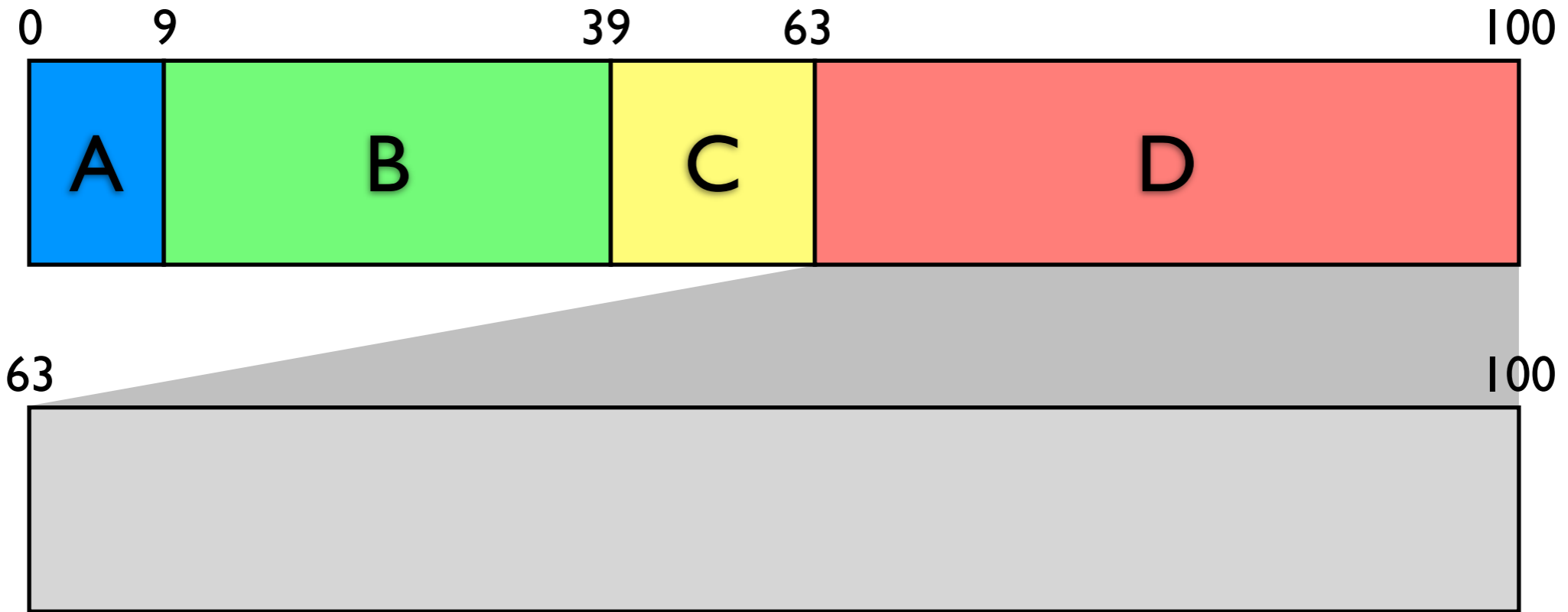
Arithmetic compression



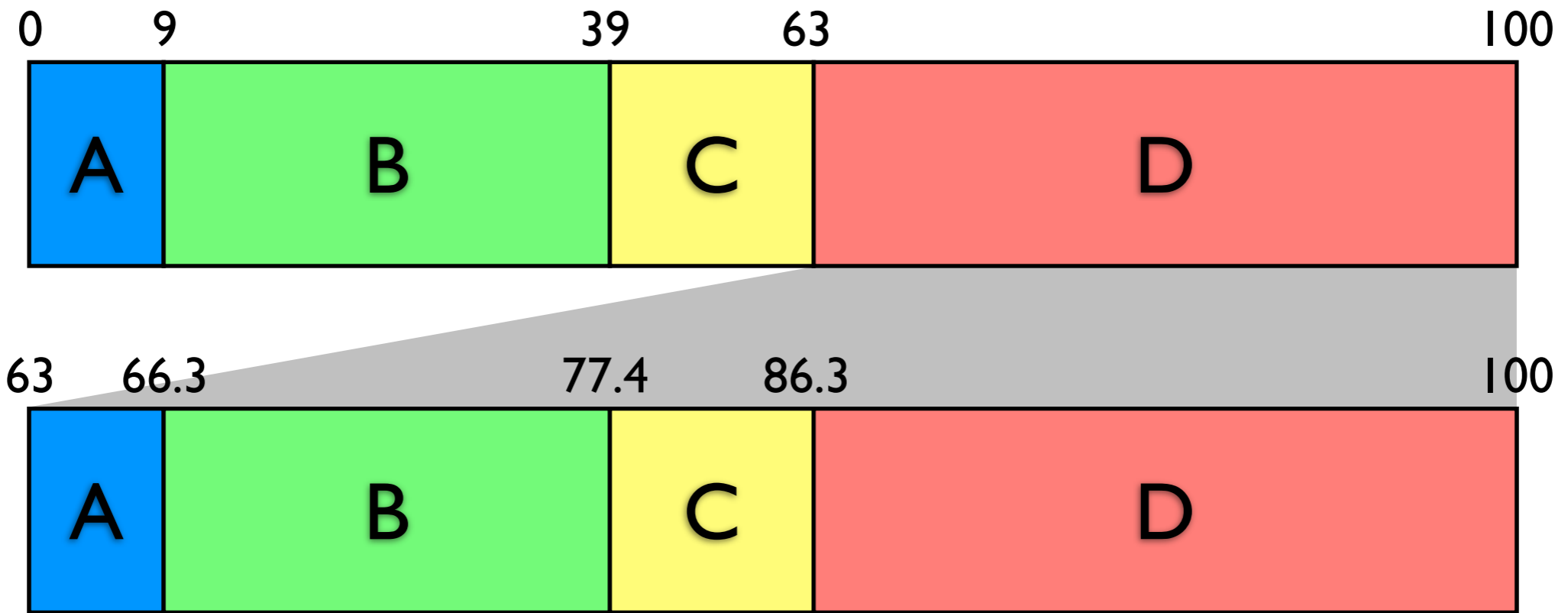
Input: D C B



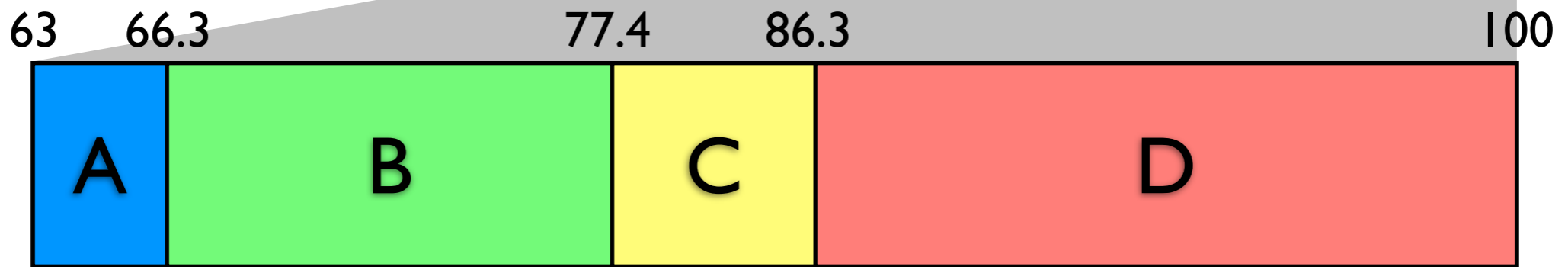
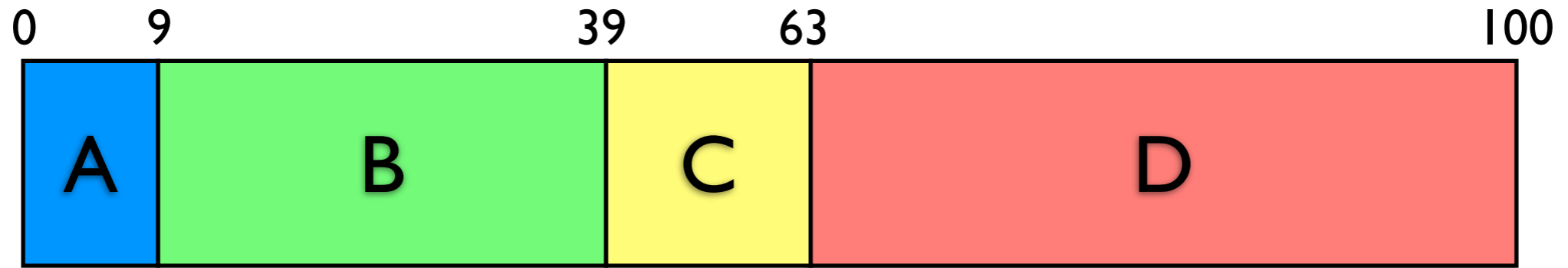
Input: D C B



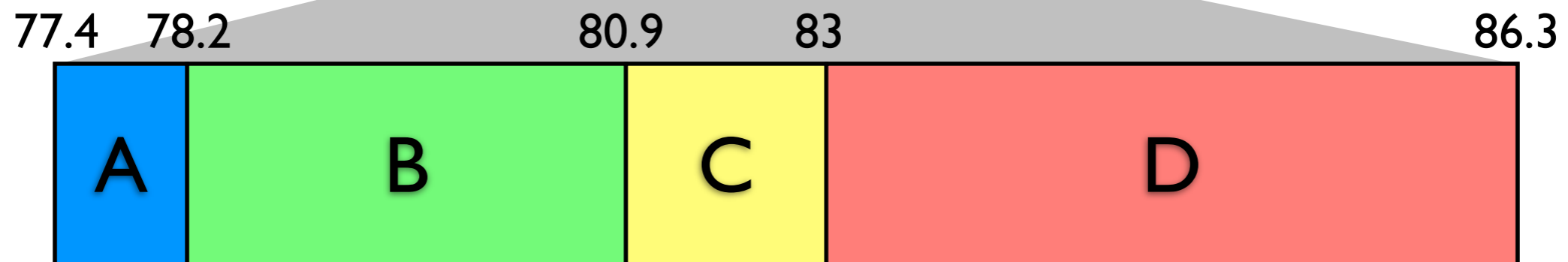
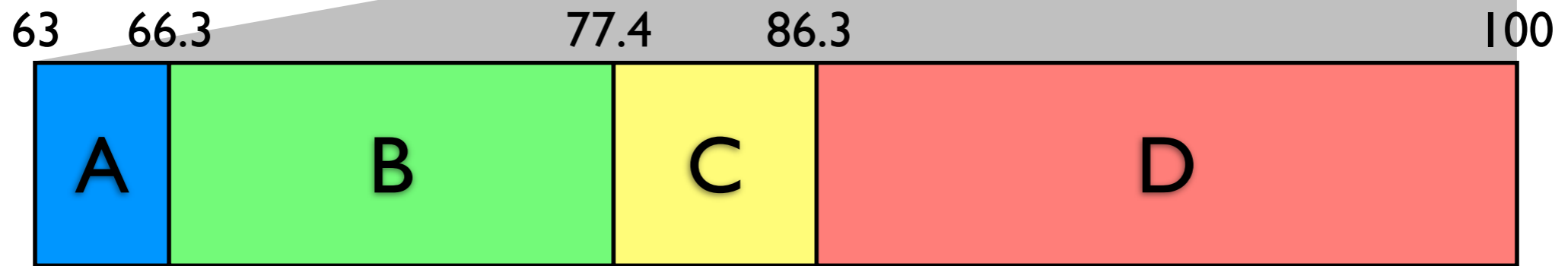
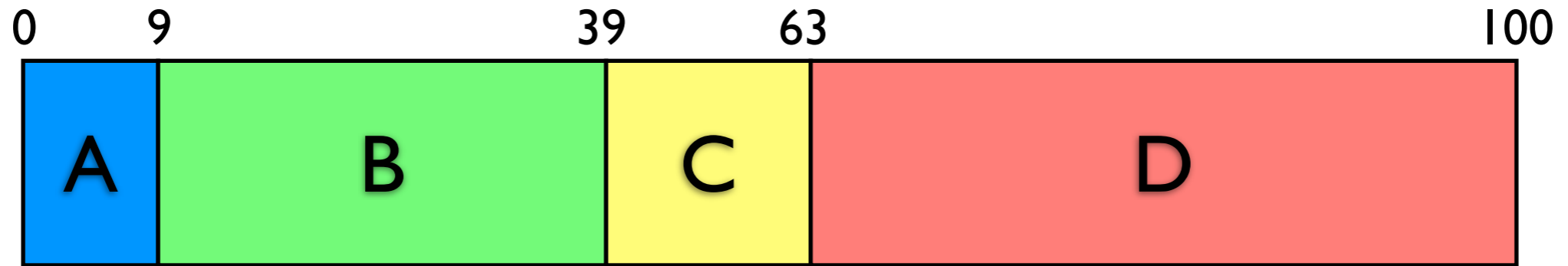
Input: D C B



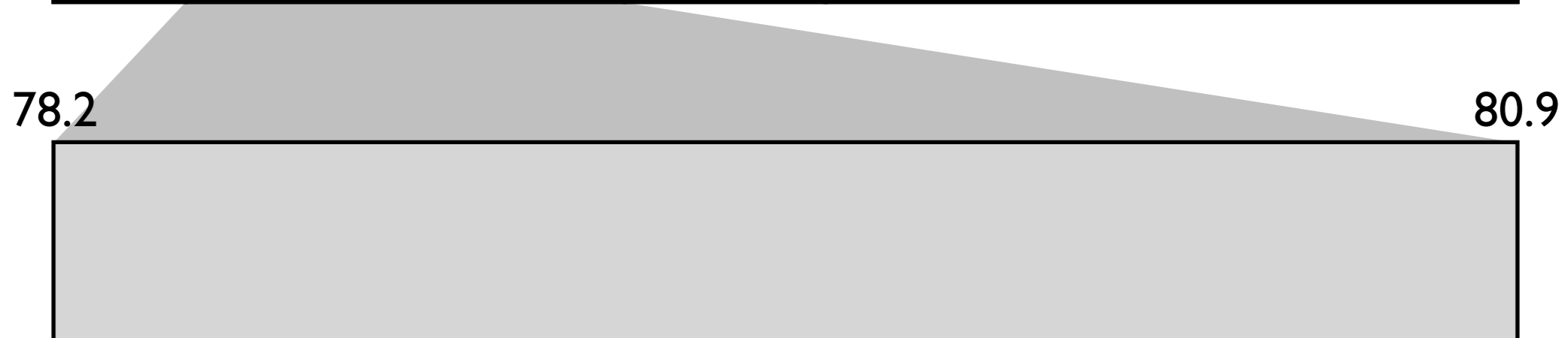
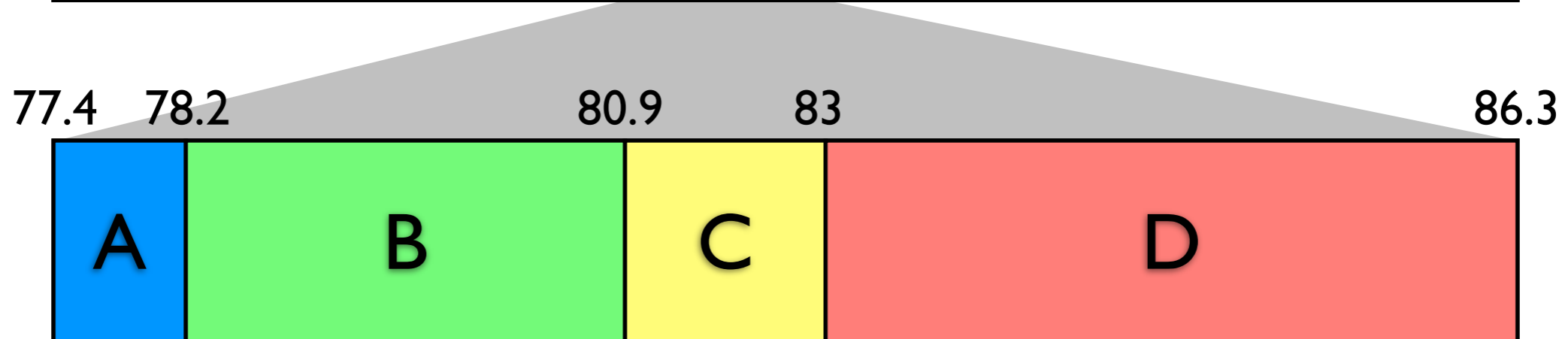
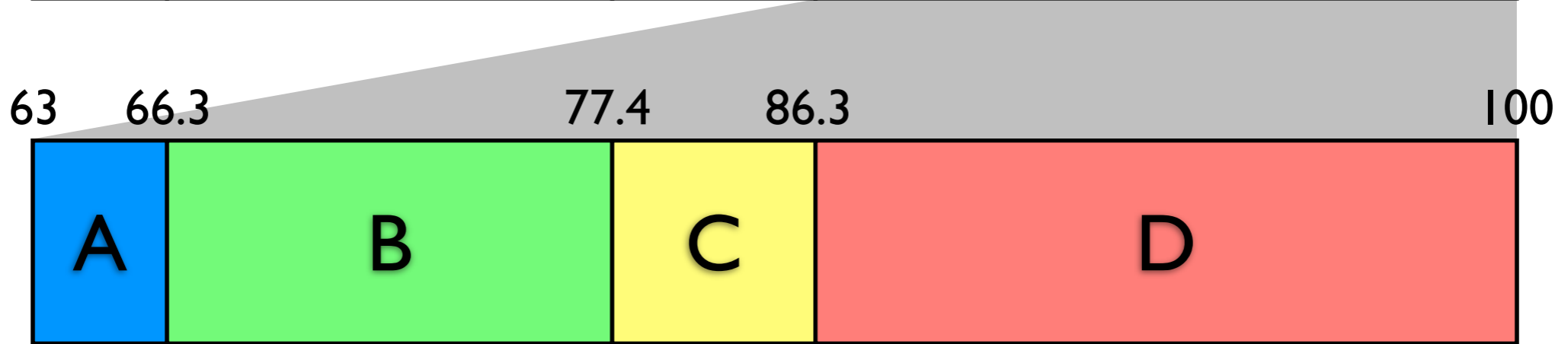
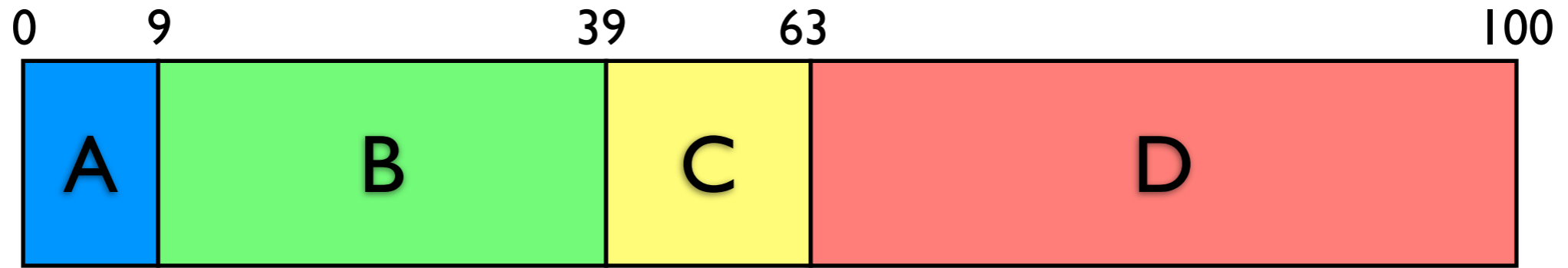
Input: D C B



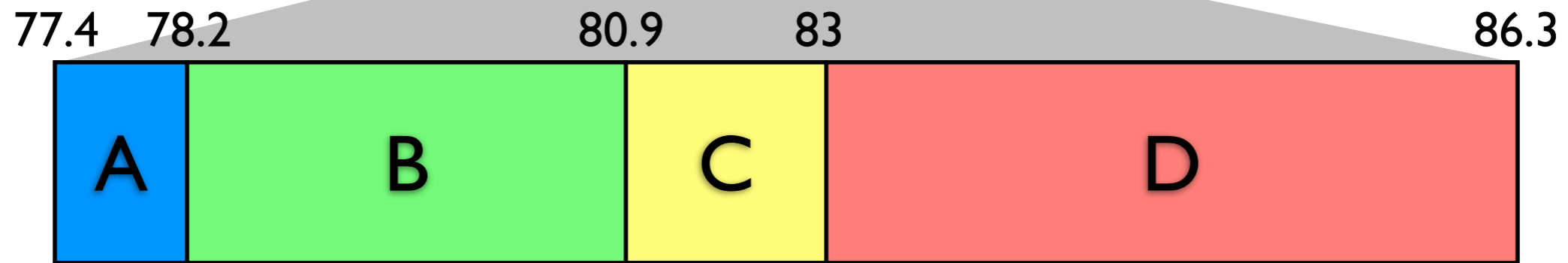
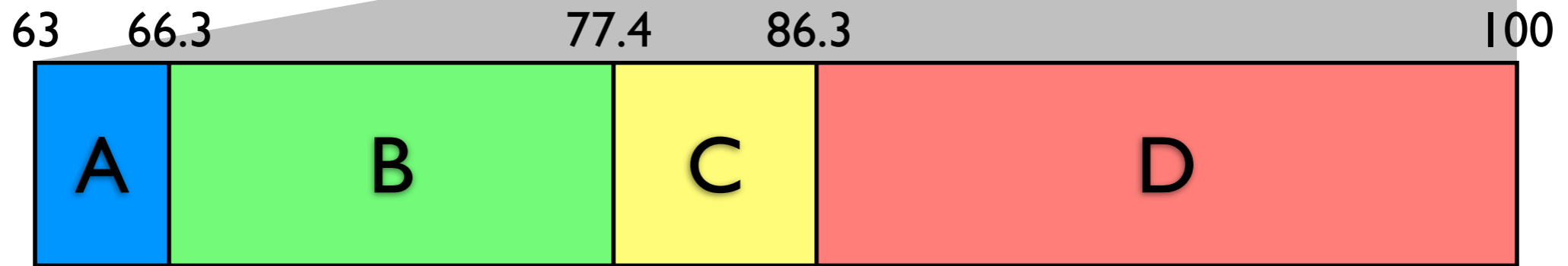
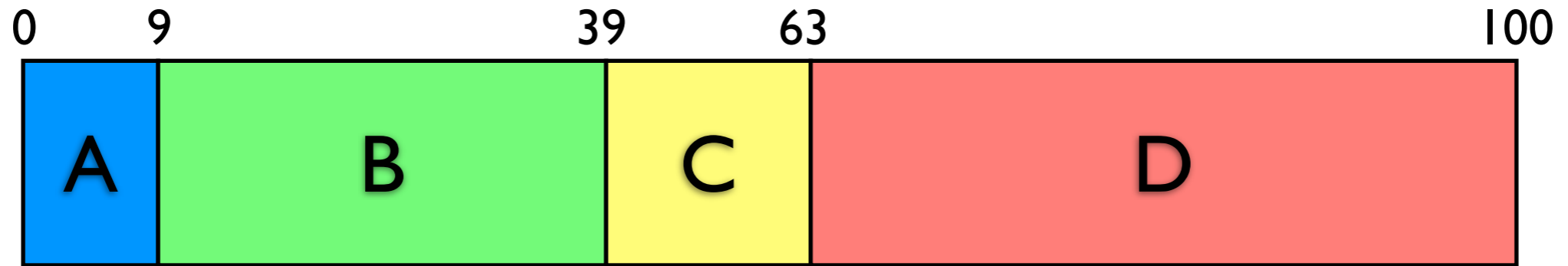
Input: D C B



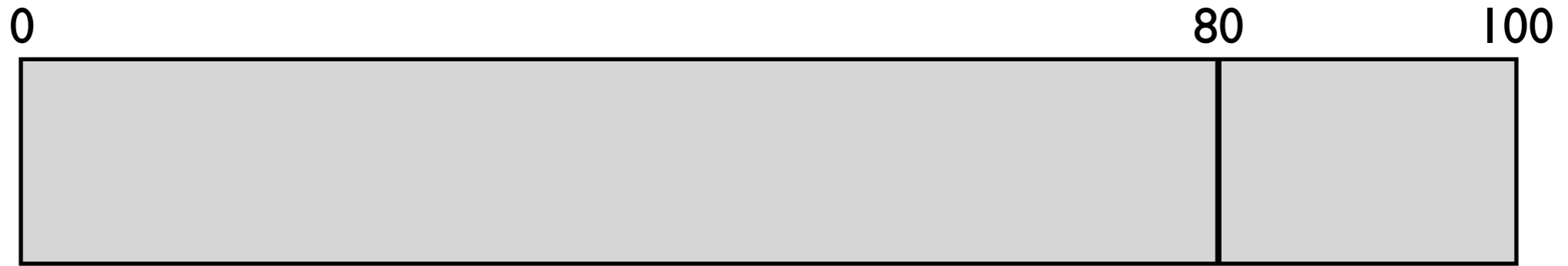
Input: D C B



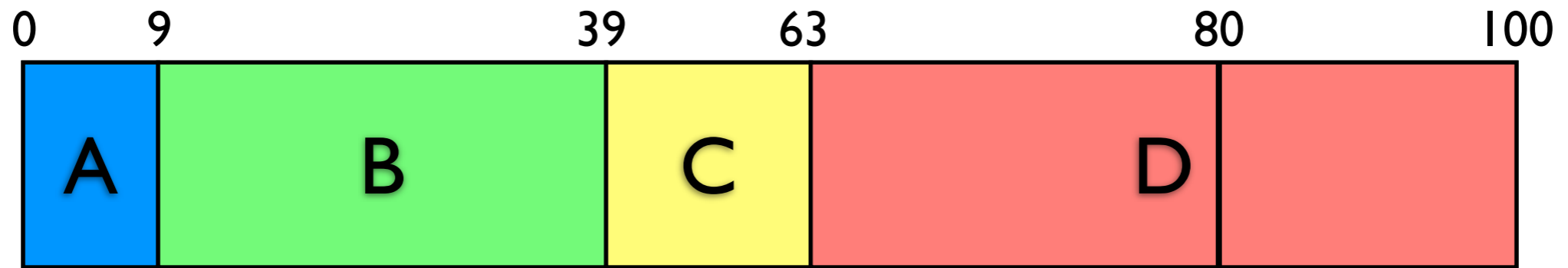
Input: D C B



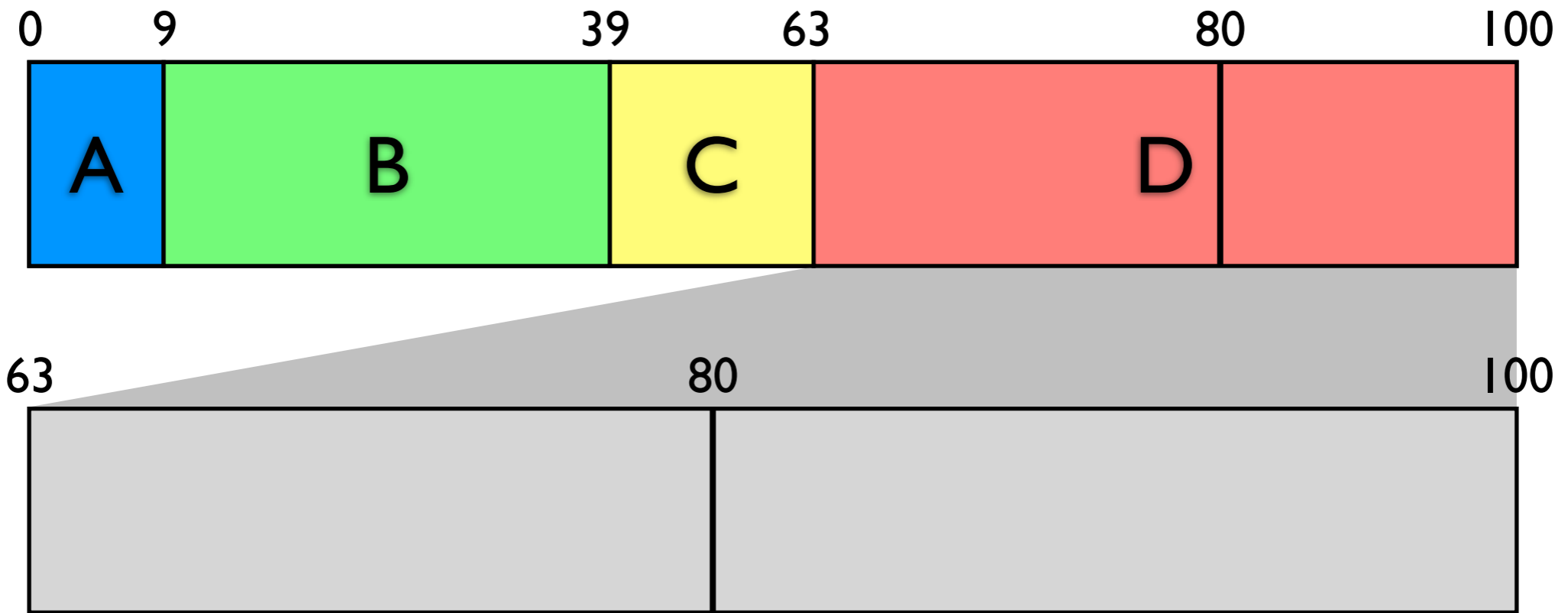
Input: 80



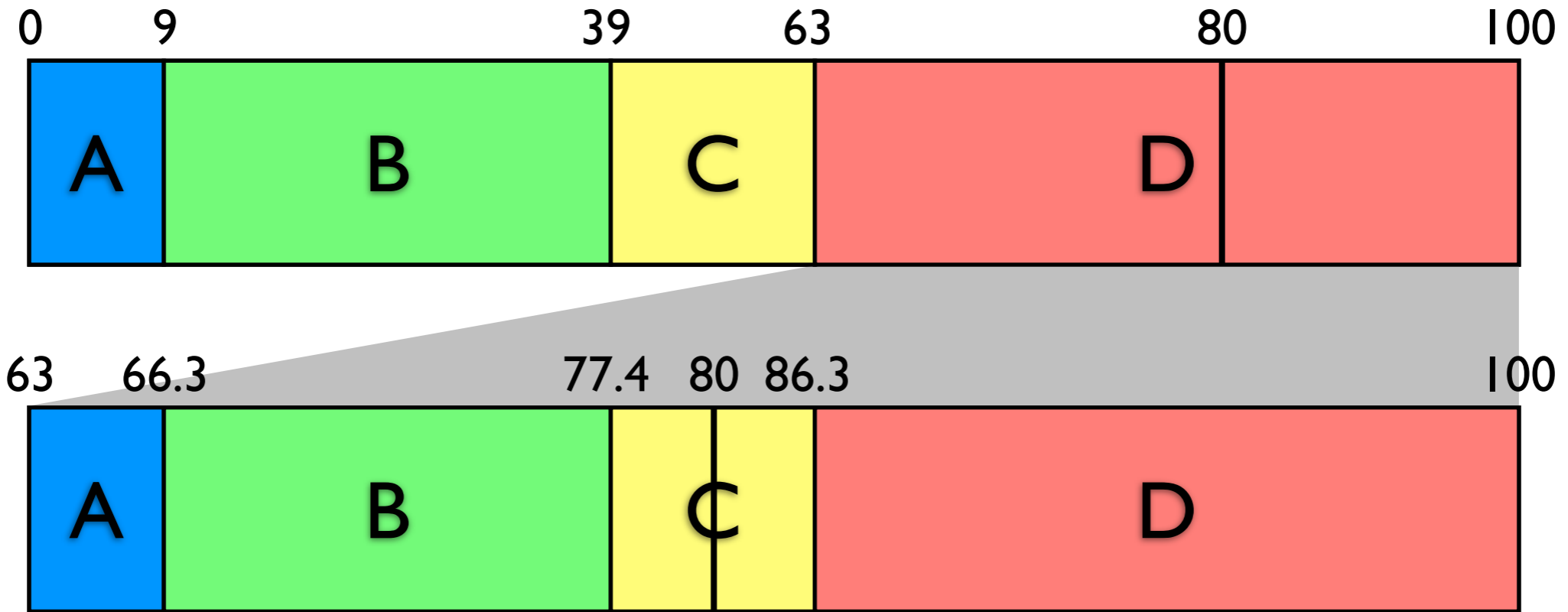
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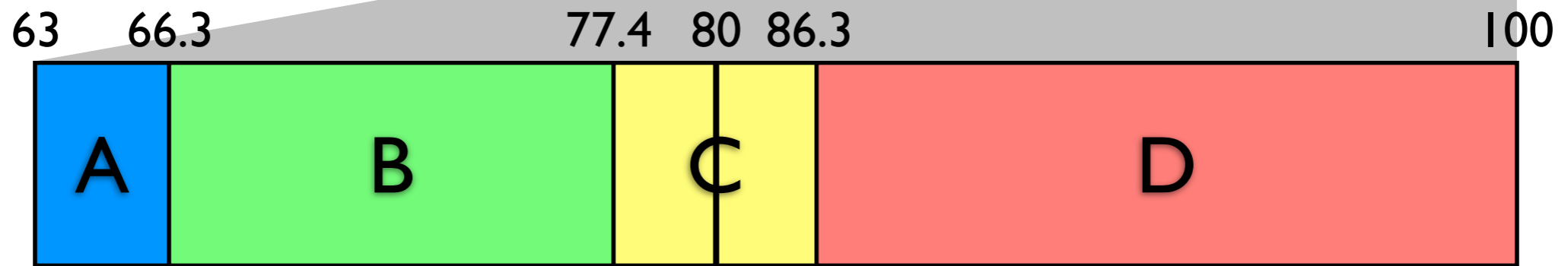
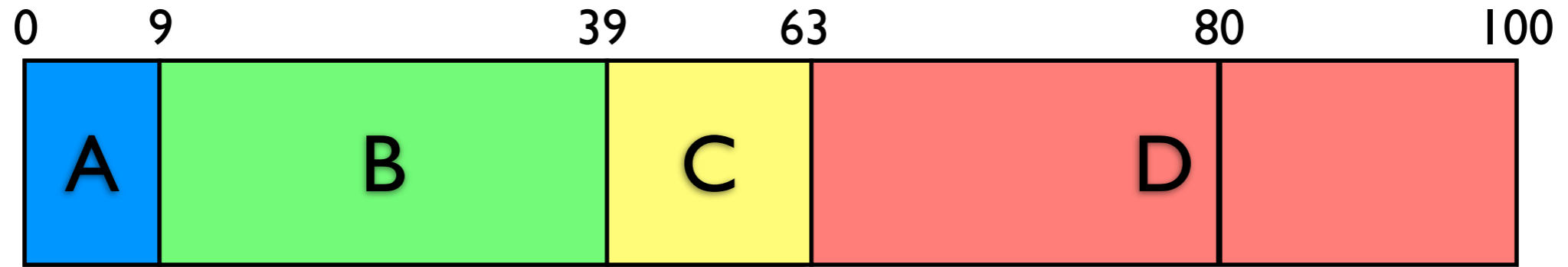
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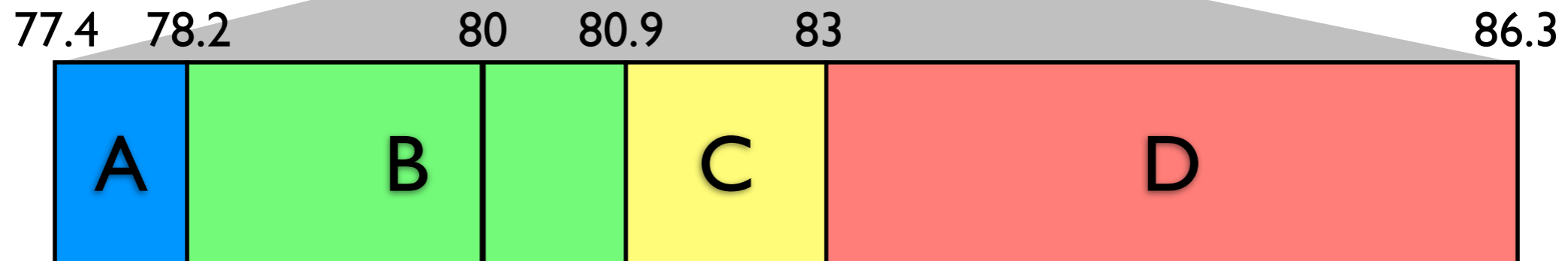
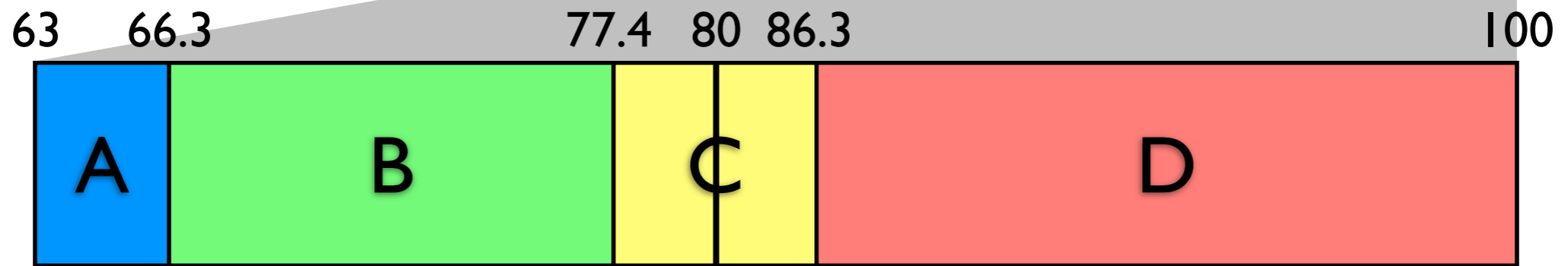
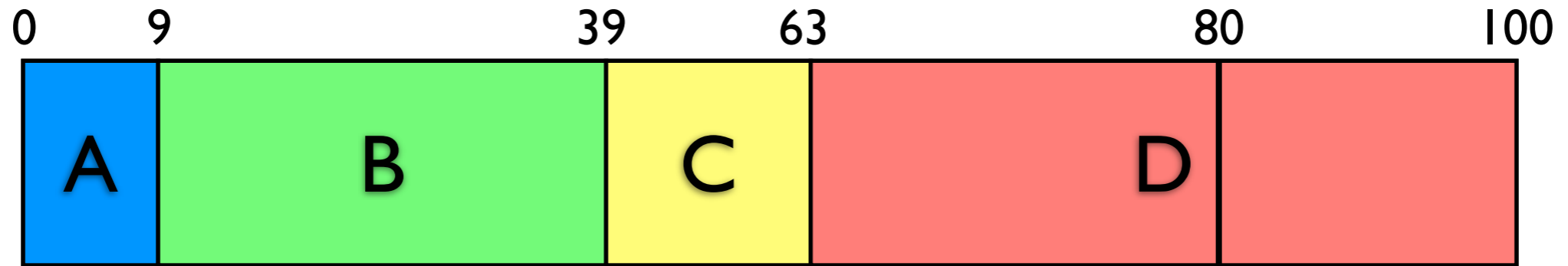
Input: 80



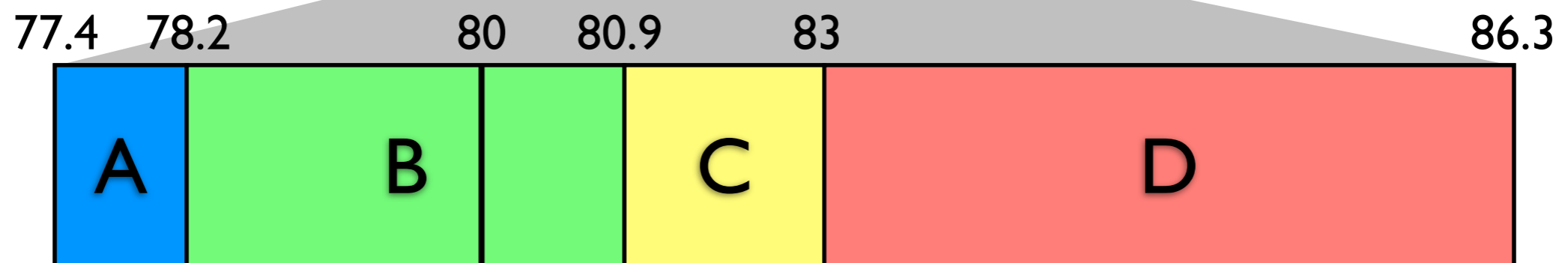
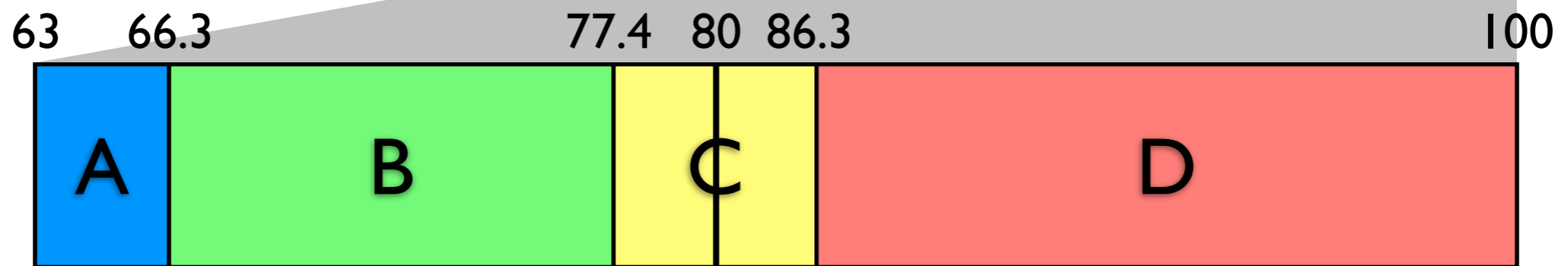
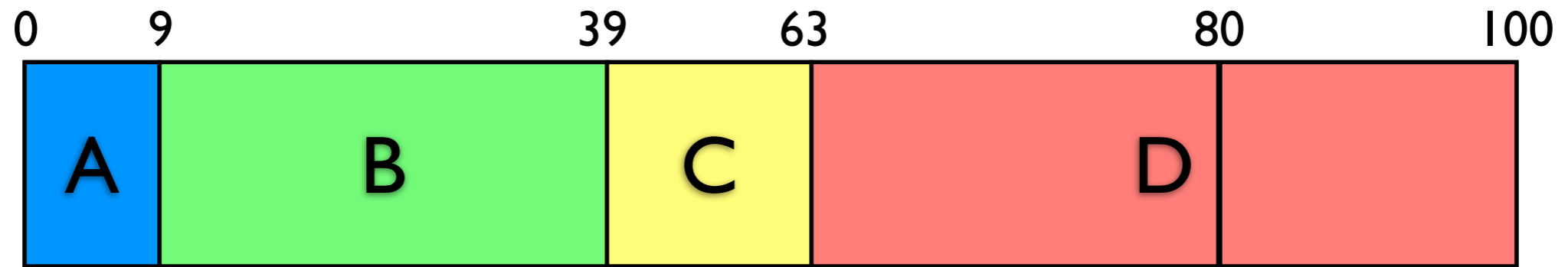
Input: 80



Input: 80



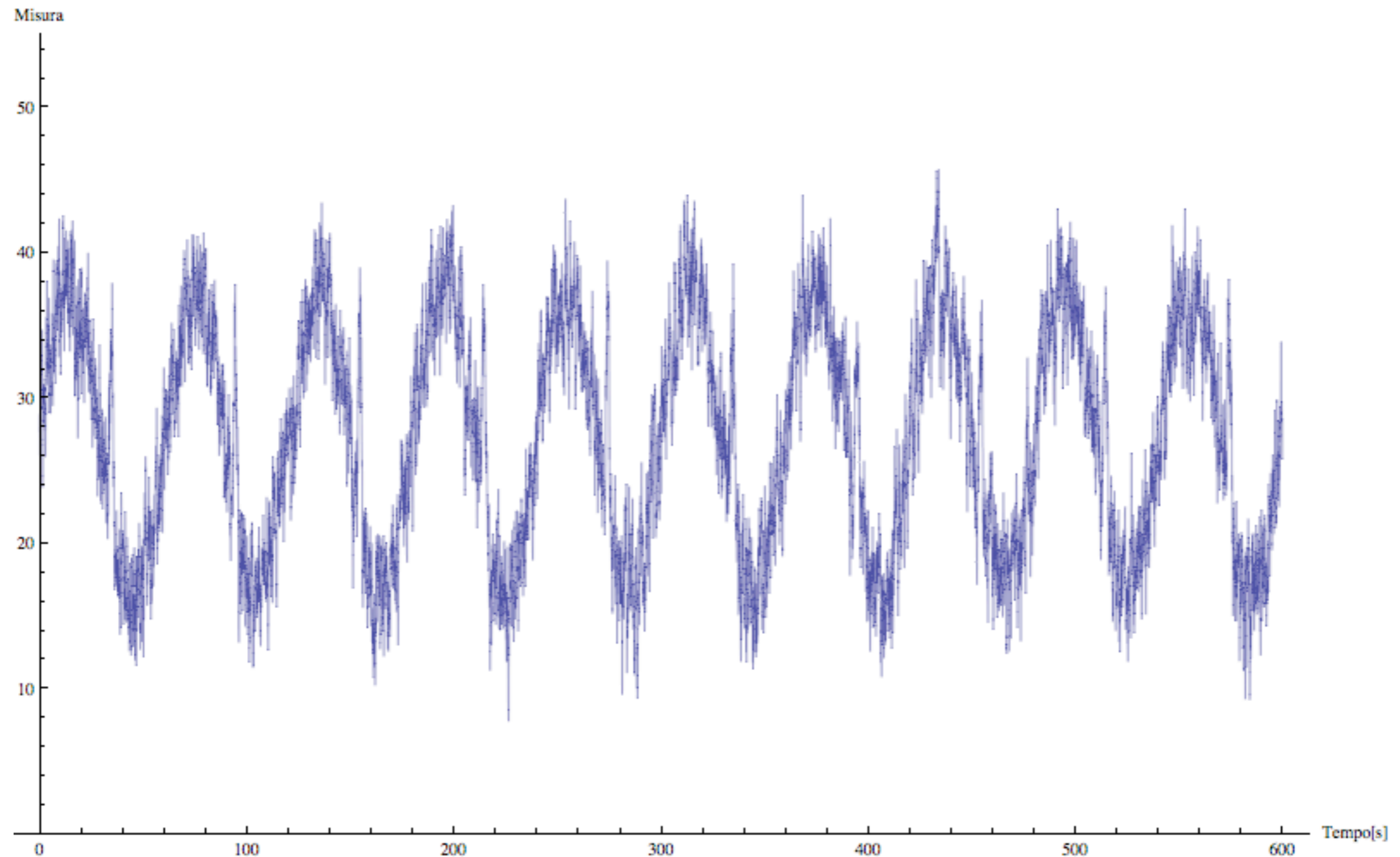
Input: 80



Part II

The Planck/LFI case

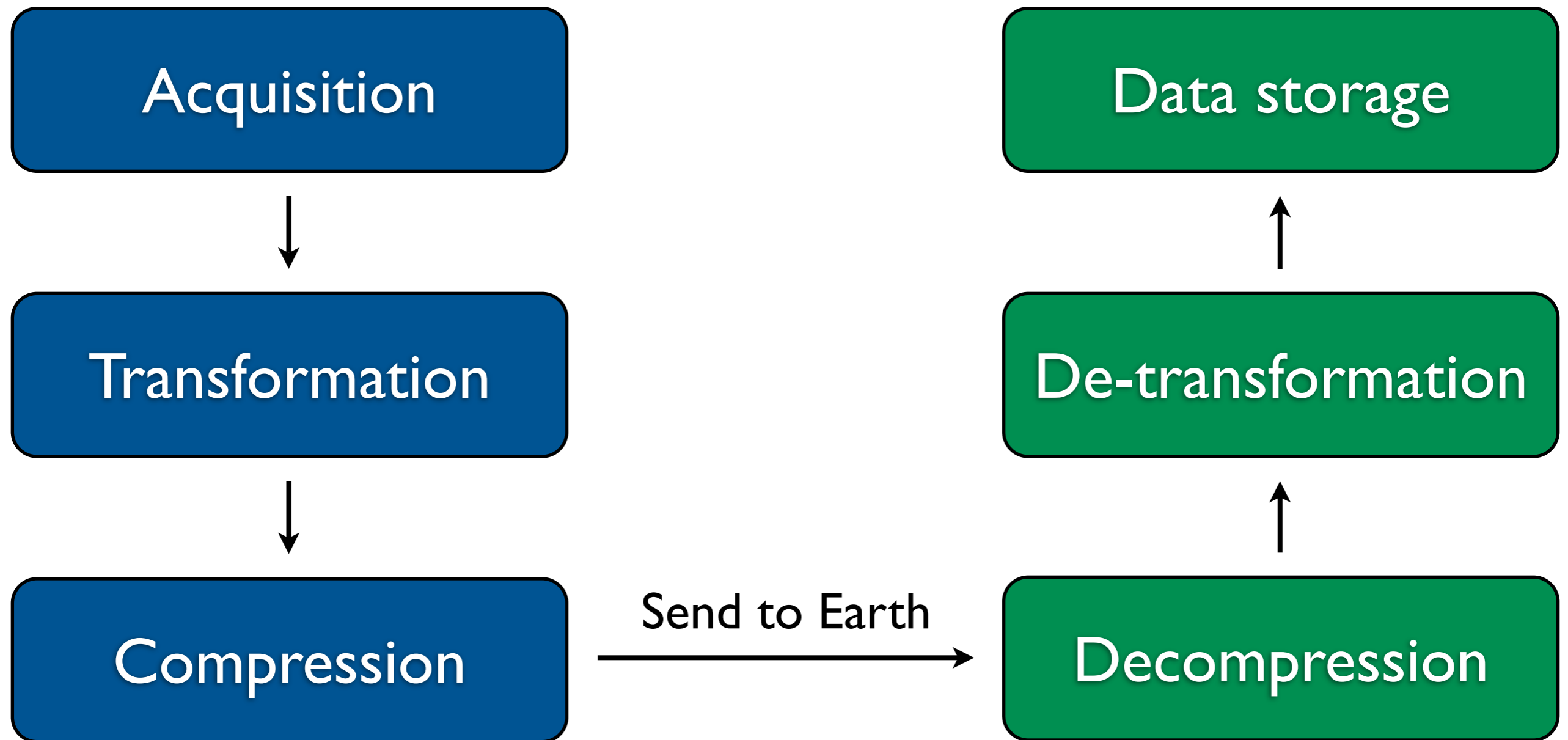
Compressing LFI data



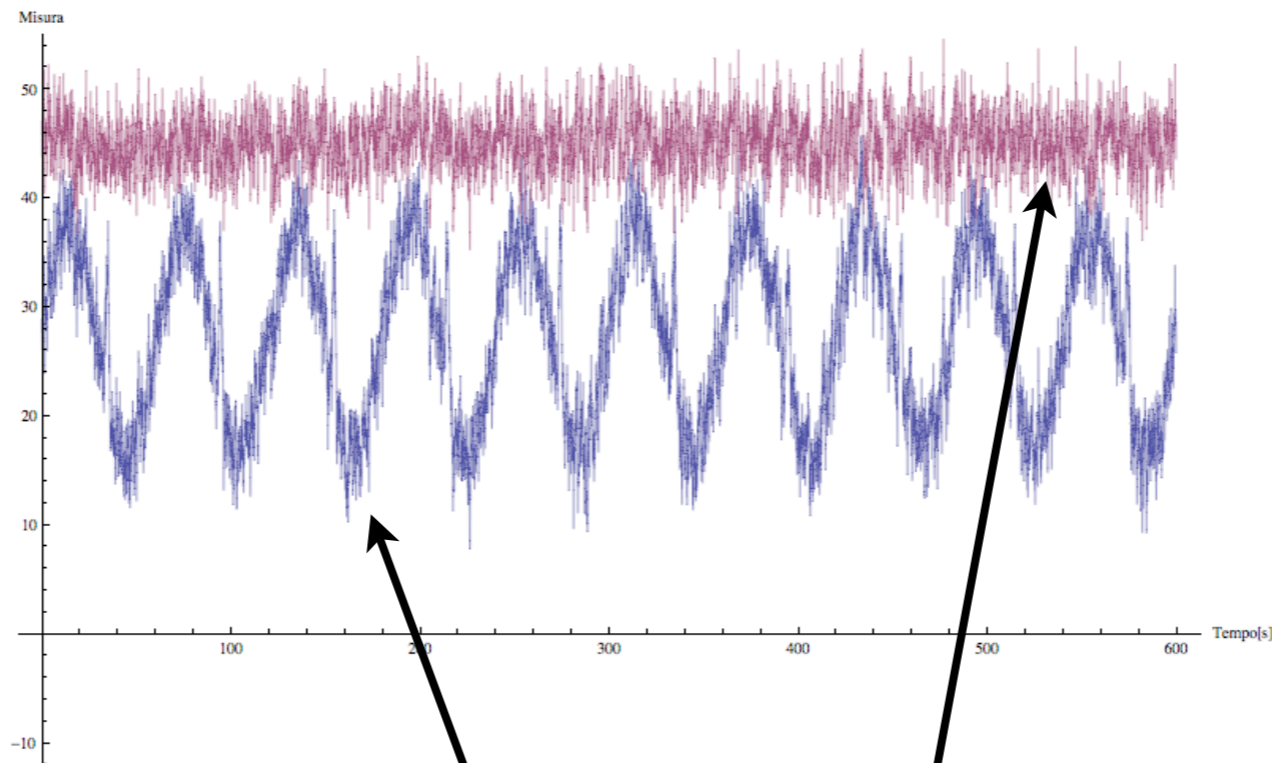
What to use with radiometric output?

- Dictionary compressors are not useful here (no recurring patterns)
- Arithmetic compression (a statistical technique) is the best choice...
- ...but it is not able to reach $c_r \sim 2.4$: we need to decrease H before the compression!

The LFI compression pipeline



A non-reversible transform



$$q_1 = \text{round}_{16} \left(\left(x_s - r_1 x_r + \Delta \right) \times s \right),$$

$$q_2 = \text{round}_{16} \left(\left(x_s - r_2 x_r + \Delta \right) \times s \right)$$

A non-reversible transform

- Differencing sky/ref reduces noise.
- The rounding operation increases the signal RMS, but in a way that *decreases H!*
- Requirement: such increment must be < 10% of the intrinsic signal RMS.

Optimisation and verification

- To find the values of the parameters r_1 , r_2 , Δ , s for each detector (44 total), LFI sent *uncompressed* data to Earth for ~ 1 day.
- We analyzed such data on ground, found the best configuration and sent it back to LFI.
- During operations, we monitor the performance of the compressor daily.

Channel #1801

TOI statistics

Time Interval	[1626455700.0, 1626458400.0] s
Number of samples	211874
V_{sky} mean \pm rms	11350.10 \pm 32.44 ADU
V_{load} mean \pm rms	11806.20 \pm 33.56 ADU
$\sigma_{\text{sky,load}}, \rho_{\text{sky,load}}$	11.9735 ADU ² , 0.0313
r	0.9614

Optimized REBA Parameters

N_{aver}	52.0000
r_1	1.0000
r_2	0.9167
\mathcal{O}	-35.82
s_q	2.9916
Quack warning	no

Processing

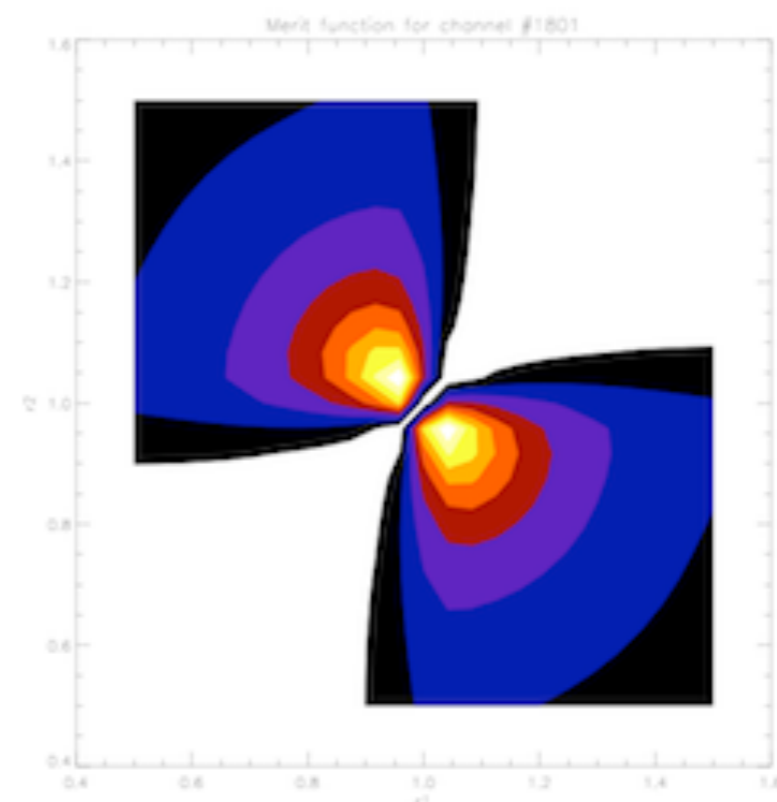
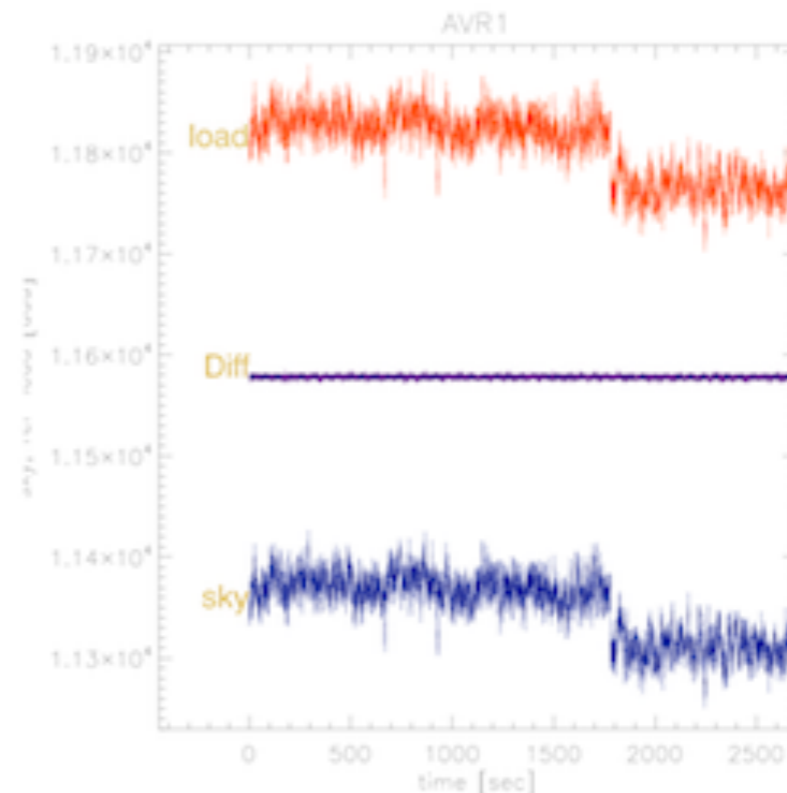
mean[P_1] + $\mathcal{O} \pm \sigma_1$	-491.99 \pm 2.34 ADU
mean[P_2] + $\mathcal{O} \pm \sigma_2$	+491.86 \pm 2.58 ADU
mean[Q_1] $\pm \sigma_{Q_1}$	-1471.83 \pm 7.02 ADU
mean[Q_2] $\pm \sigma_{Q_2}$	+1471.43 \pm 7.74 ADU
max(Q_{ack})	0.045929
C_r	2.37 \pm 0.03 (target: 2.40)
η_{Cr}	(82.49 \pm 0.49)%
H_{pck}	5.57 \pm 0.05 bits
H_{toi}	5.93 bits
ϵ_q	1.5699 ADU (sky), 1.6366 ADU (ref), 0.0684 ADU (diff.)
$\frac{\epsilon_q}{\sigma}$	0.137859 (sky), 0.048750 (ref), 0.035396 (diff.)

Telemetry

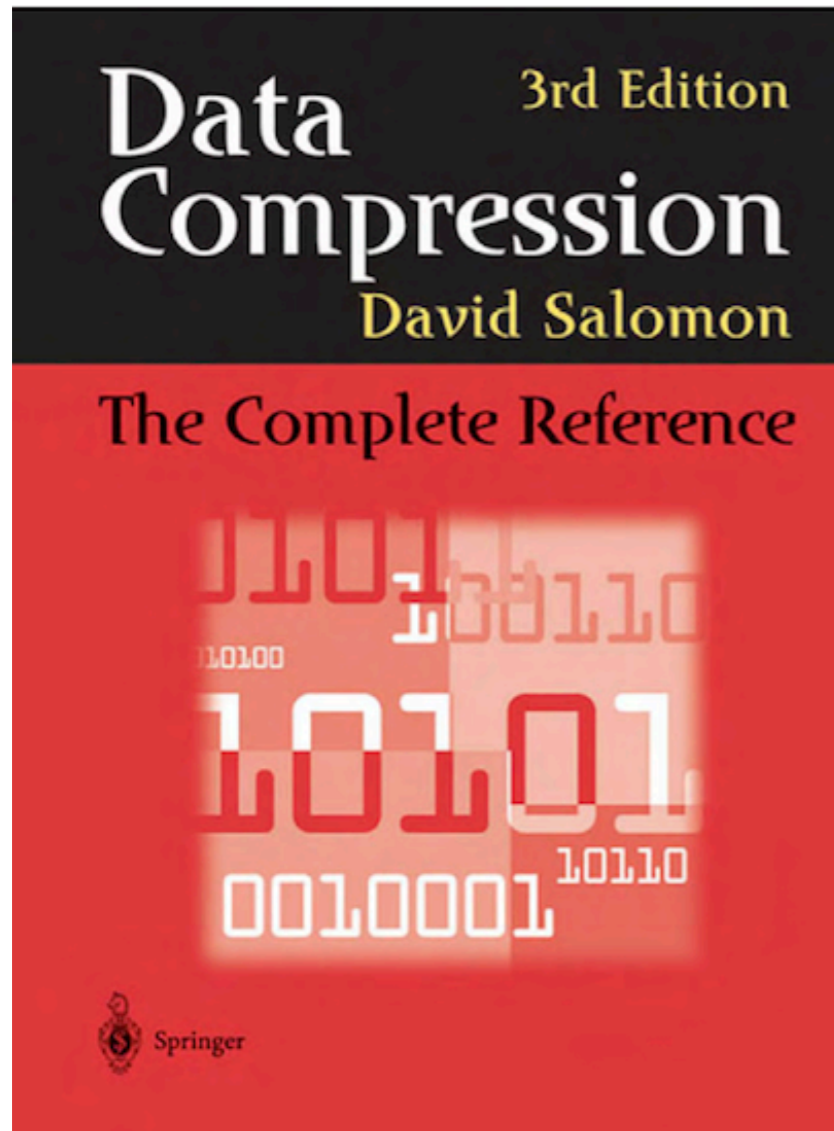
N_{pck}	414
Uncomp. data length	847496. bytes
Comp. data length	356900 bytes
Comp. data rate	1111.45 bits/sec

Details of the analysis

Algorithm	iterative ('diff' strategy)
Grid $r_1 \times r_2$	[0.50, 1.50] \times [0.50, 1.50] (in 25 \times 25 steps)
Elapsed Time	60.5 s



Additional references



*J*inst

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THE PLANCK LOW FREQUENCY INSTRUMENT

Optimization of Planck-LFI on-board data handling

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