

## A. Learning Better Lossless Compression Using Lossy Compression – Supplementary

### A.1. Q-Classifier Architecture

We show the architecture for the Q-Classifier in Table A1. *Residual* denotes a sequence of convolution, ReLU, convolution, with a skip connection adding the input to the output (as in [15], but without BatchNorm).

Layer	$C_{\text{in}}$	$C_{\text{out}}$	Filter	Stride
Conv + ReLU	3	64	$5 \times 5$	2
Conv + ReLU	64	128	$5 \times 5$	2
$4 \times$ Residual	128	128	$3 \times 3$	
Conv	128	256	$5 \times 5$	2
$4 \times$ Residual	256	256	$3 \times 3$	
Channel-Avg.	256	256		
Linear	256	$ \mathcal{Q} $		

Table A1. Q-Classifier architecture.

### A.2. BPG Performance

Fig. A1 compares the performance of BPG on Kodak, in terms of PSNR, to the recent learned image compression approach from Minnen *et al.* [30]. The plot is digitized from Figure 2 in [30].

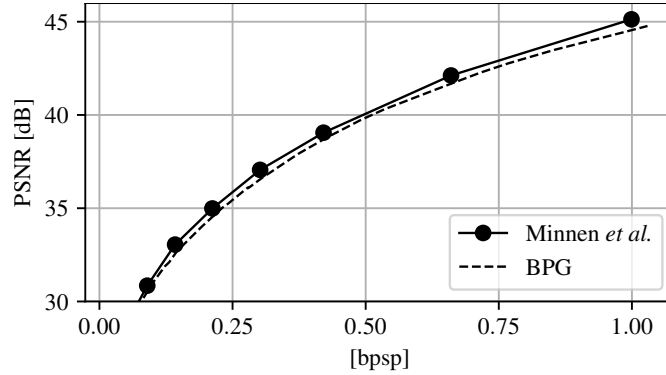


Figure A1. Comparing BPG to Minnen *et al.* [30]

### A.3. Examples from the testing sets

We provide additional visual examples here:

[https://data.vision.ee.ethz.ch/mentzerf/rc/rc\\_suppl\\_additional.pdf](https://data.vision.ee.ethz.ch/mentzerf/rc/rc_suppl_additional.pdf)

Specifically, we show one image from each of our testing sets, alongside with the residual  $r$  and a sample from  $p(r)$ , which is expected to be visually similar to  $r$ . Please refer to Section 6.4 for details on sampling and the visualization.