Network Architectures and Hyperparameters

Operation	Kernel	Strides	Feature Maps	BN	Nonlinearity			
G(z,c) - 1 x 1 x 28 input								
Dense			3136	Yes	ELU			
Reshape - $7 \times 7 \times 64$								
Transposed Convolution	4×4	2×2	128	Yes	ELU			
Transposed Convolution	4×4	1×1	64	Yes	ELU			
Transposed Convolution	4 x 4	2 x 2	1	No	Sigmoid			
$E(x) - 28 \times 28 \times 1 \text{ input}$								
Convolution	3×3	1×1	32	Yes	ELU			
Convolution	3×3	2×2	64	Yes	ELU			
Convolution	3×3	2×2	128	Yes	ELU			
Dense			1024	Yes	ELU			
Output z: Dense			1	No	Tanh			
Output $c_{\rm disc}$: Dense			10	No	Softmax			
Output c_{cont} : Dense			2	No	Linear			
D(x) - 28 x 28 x 1 input								
Convolution	3×3	2×2	64	Yes	ELU			
Convolution	3×3	2×2	128	Yes	ELU			
Dense			512	Yes	ELU			
D(Z) - 1 x 1 x 28 input								
Convolution	1×1	1×1	64	Yes	ELU			
Convolution	1×1	1×1	128	Yes	ELU			
Dense			512	Yes	ELU			
D(x,Z) - 1 x 1 x 1024 input								
concatenate $D(x)$ and $D(Z)$ along channel axis								
Dense			1024	Yes	ELU			
Dense			1	No	Sigmoid			
Optimizer D:	Adam ($\alpha = 10^{-4},$	$\beta_1 = 0.5$)					
Optimizer G and E:	Adam ($\alpha = 5 \times 10^{-4}, \beta_1 = 0.5$)							
Learning Rate Reduction	Polynomial with final LR one fifth of the original							
Batch Size	128							
Iterations	30 000							
G: weight, bias init	Truncated Normal ($\mu = 0, \sigma = 0.02$)							
D and E: weight, bias init	$\mathcal{N}(0,\sqrt{inp/2}^{-1})$ with inp number of inputs to a unit							

 ${\bf Table~1:~MNIST~Model~Hyperparameters}$

Operation	Kernel	Strides	Feature Maps	BN	Nonlinearity			
G(z,c) - 1 x 1 x 98 input								
Dense			2048	Yes	ELU			
Reshape - $2 \times 2 \times 512$								
Transposed Convolution	5×5	2×2	512	Yes	ELU			
Transposed Convolution	5×5	2×2	256	Yes	ELU			
Transposed Convolution	5×5	1×1	128	Yes	ELU			
Transposed Convolution	5×5	1×1	64	Yes	ELU			
Transposed Convolution	5×5	1×1	32	Yes	ELU			
Transposed Convolution	4×4	1×1	3	No	Sigmoid			
$E(x) - 32 \times 32 \times 3 \text{ input}$								
Convolution	3×3	1×1	32	Yes	ELU			
Convolution	3×3	2×2	64	Yes	ELU			
Convolution	3×3	1×1	128	Yes	ELU			
Convolution	3×3	2×2	256	Yes	ELU			
Convolution	3×3	2×2	512	Yes	ELU			
Dense			1024	Yes	ELU			
Output z: Dense			1	No	Tanh			
Output $c_{\rm disc}$: Dense			4×10	No	Softmax			
Output c_{cont} : Dense			4	No	Linear			
D(x) - 32 x 32 x 3 input								
Convolution	4×4	2×2	64	Yes	ELU			
Convolution	4×4	2×2	128	Yes	ELU			
Convolution	4×4	2×2	256	Yes	ELU			
Dense			1024	Yes	ELU			
D(Z) - 1 x 1 x 28 input								
Convolution	1×1	1×1	64	Yes	ELU			
Convolution	1×1	1×1	128	Yes	ELU			
Convolution	1×1	1×1	256	Yes	ELU			
Dense			1024	Yes	ELU			
D(x,Z) - 1 x 1 x 2048 input								
concatenate D(x) and D(Z) along channel axis								
Dense			1024	Yes	ELU			
Dense			1	No	Sigmoid			
Optimizer D, G, and E:	Adam ($\alpha = 10^{-4}$	$\beta_1 = 0.5$)					
Learning Rate Reduction	Polynomial with final LR one fifth of the original							
Batch Size	32							
Iterations	300 000 (SVHN); 500 000 (CelebA							
G: weight, bias init	Truncated Normal ($\mu = 0, \sigma = 0.02$)							
D and E: weight, bias init	$\mathcal{N}(0,\sqrt{inp/2}^{-1})$ with inp number of inputs to a unit							
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Table 2: CelebA / SVHN Model Hyperparameters