



AMD

Neural Texture Block Compression

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Motivation

- Texture is a key component to achieve high visual fidelity through material properties
 - High-resolution textures (e.g., 4K) require a lot of storage
- Block compression (BC) is one of the most popular techniques to compress textures
 - All variations, BC1 BC7, compress each 4x4 texel block to a fixed number of bytes
 - BC1 and BC4: 8 bytes, others: 16 bytes
- Consumes 8MB (= 4096*4096 / (4*4) * 8 bytes) for a single 4K texture with BC1

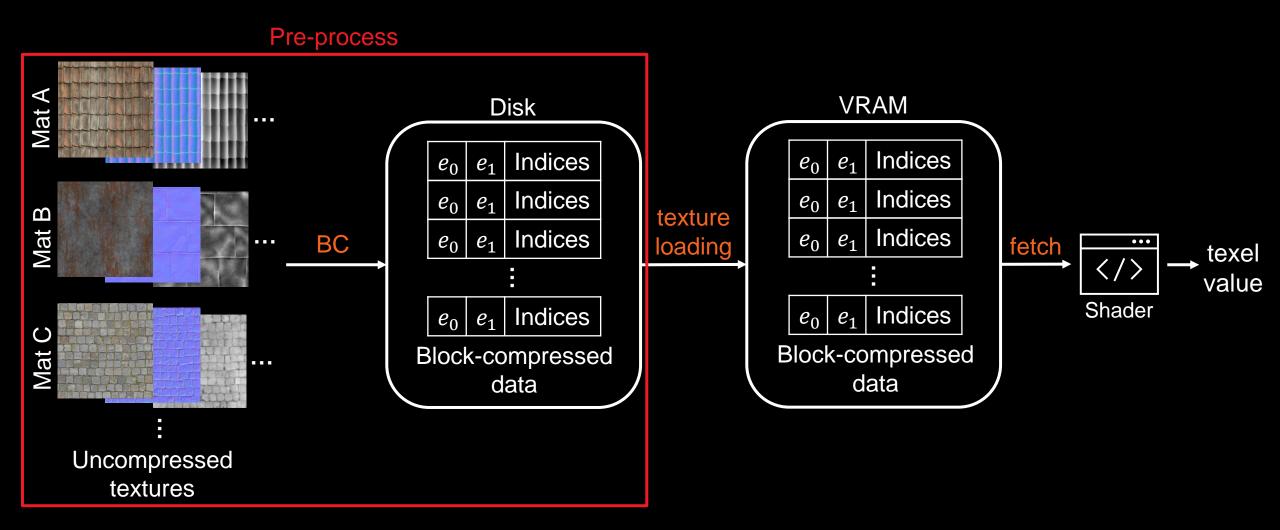
- To reduce the storage costs, we propose Neural Texture Block Compression (NTBC)
 - Compress textures in BC1/BC4 formats using a multi-layer perceptron (MLP)
 - Not require any change in the shader execution







Pipeline - conventional

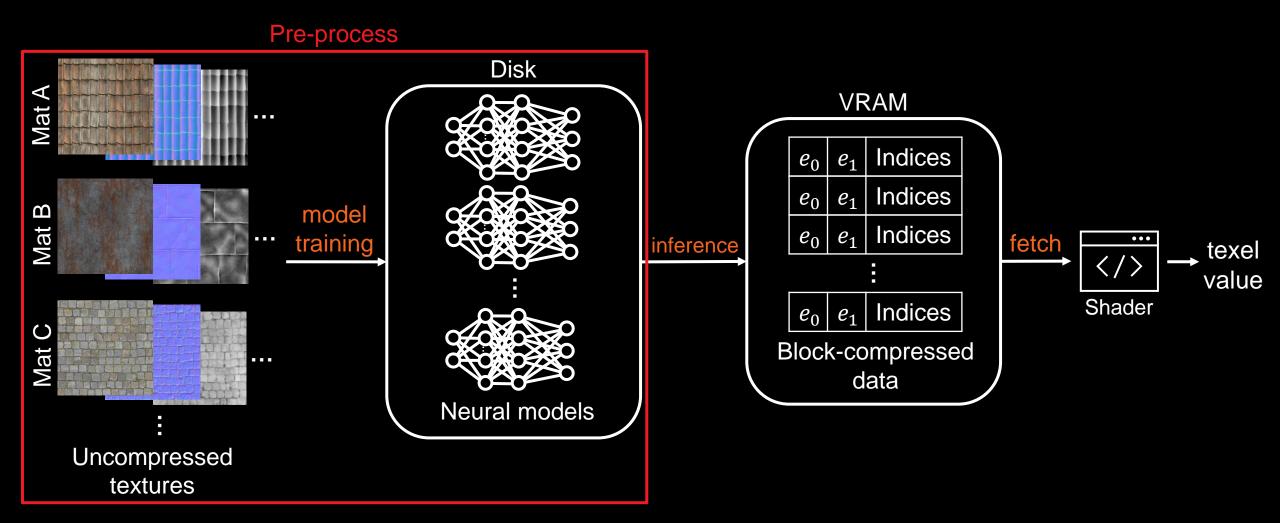






NER

Pipeline - NTBC



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Block Compression

- Block compression (BC) encodes 4x4 texel blocks into a fixed-size structure
- Each block contains a color palette with colors on a line segment in RGB space
 - Two endpoints + linear-interpolations

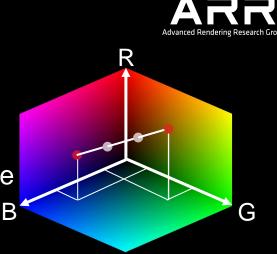
<u>BC1</u>

- RGB images
- 2 RGB565 endpoints (4 bytes)
 - Palette has 4 entries
- 16 2-bit indices (4 bytes)
 - $0 \le n \le 3$

<u>BC4</u>

- Single-channel images
- 2 8-bit endpoints (2 bytes)
 - Palette has 8 entries
- 16 3-bit indices (6 bytes)
 - $0 \le n \le 7$

How can we encode these block-compressed textures using neural networks?





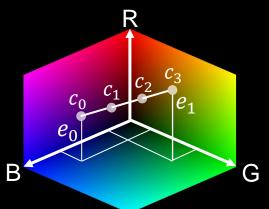
Naive approach

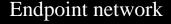
• Colors of the palette are evenly spaced on a line segment in RGB space:

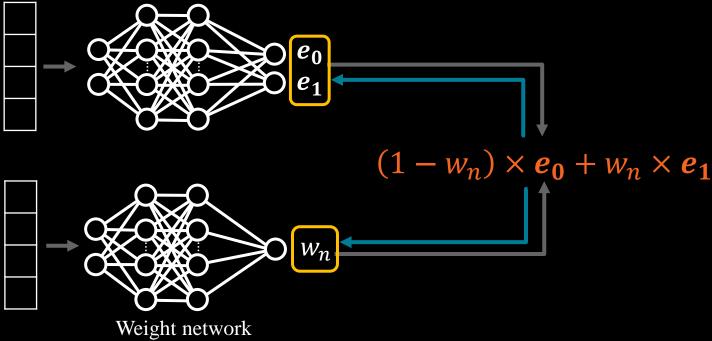
 $\boldsymbol{c_n} = (1 - w_n) \times \boldsymbol{e_0} + w_n \times \boldsymbol{e_1}$

 e_0, e_1 : two endpoints, w_n : weights of $\frac{n}{3}$ (BC1), $\frac{n}{7}$ (BC4)

Encoding e_0, e_1, w_n using NNs should be a straight-forward approach

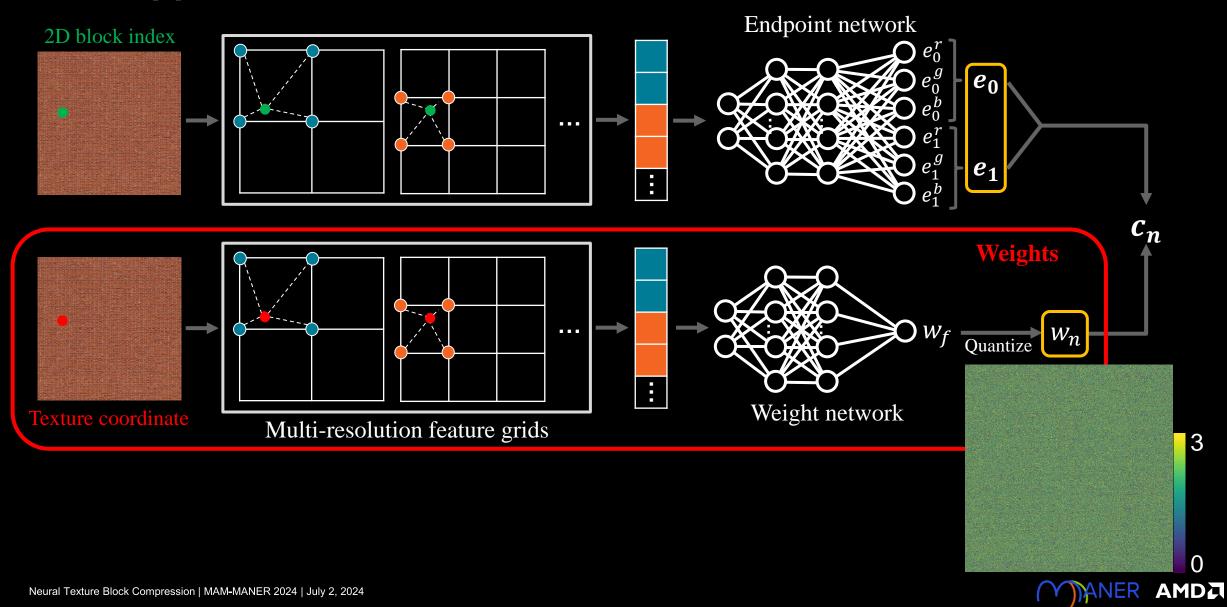






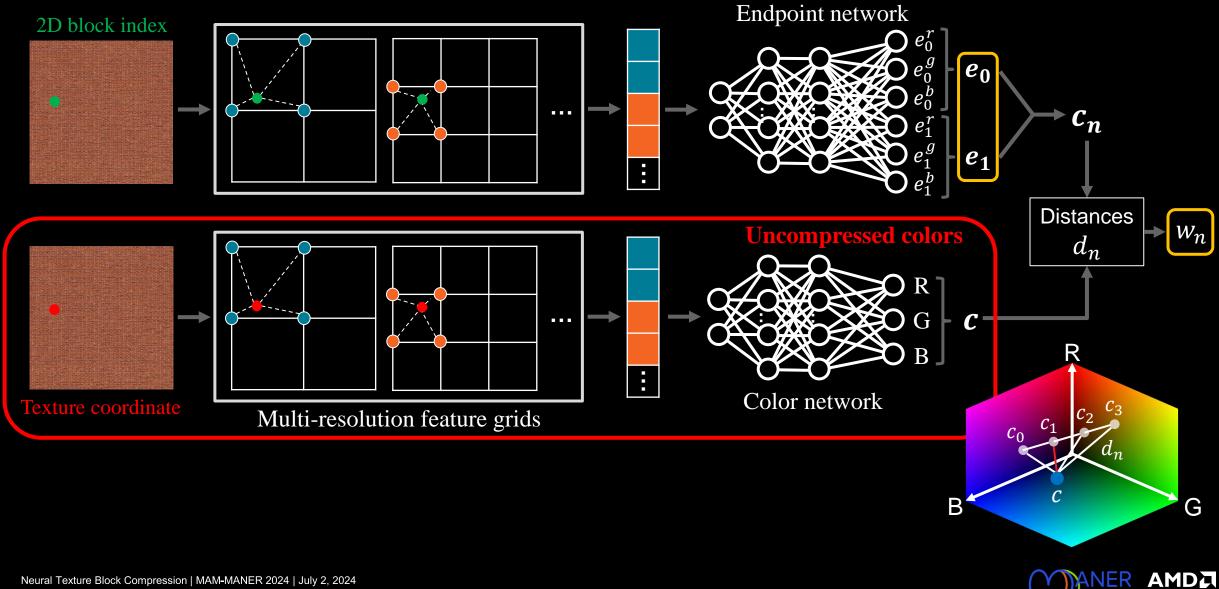


Naive approach





Neural Texture Block Compression



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NTBC – model configuration

- Multi-resolution feature grids <u>Endpoint network</u>
 - 7 levels

R

- Resolutions from 16 to 1024
- 2D features per level
- Color network
- 8 levels
- Resolutions from 16 to 2048
- 2D features per level
- Small MLPs with FP16
 - 64 x 3 hidden layers
- Adam optimizer



NTBC – grid quantization

- Multi-resolution feature grids
 <u>Endpoint network</u>
 FP16 7 levels
 FP16
 - Resolutions from 16 to 1024
 5.33 MB
 - 2D features per level
 <u>Color network</u>
 - 8 levels

R

- Resolutions from 16 to 2048
 21.33 MB
- 2D features per level

26.67 MB

13.33 MB

26.67 MB

FP8

2.67 MB

10.67 MB

Π

Quantization

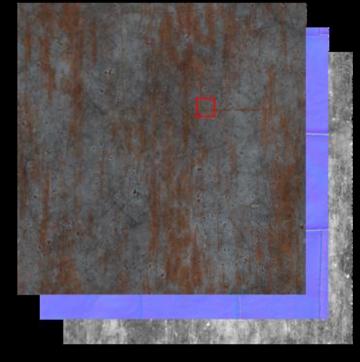
 \times 2 for RGB and single-channel textures

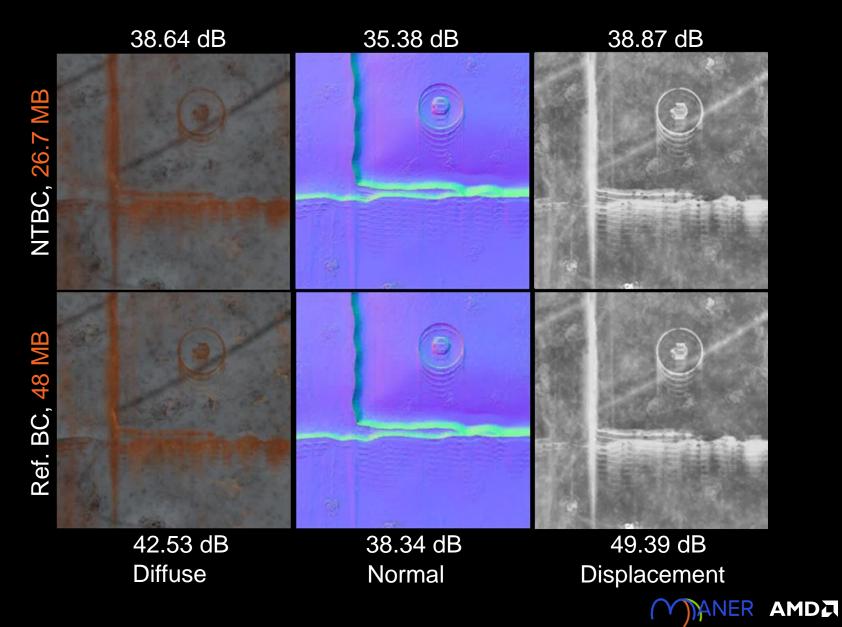




Comparisons

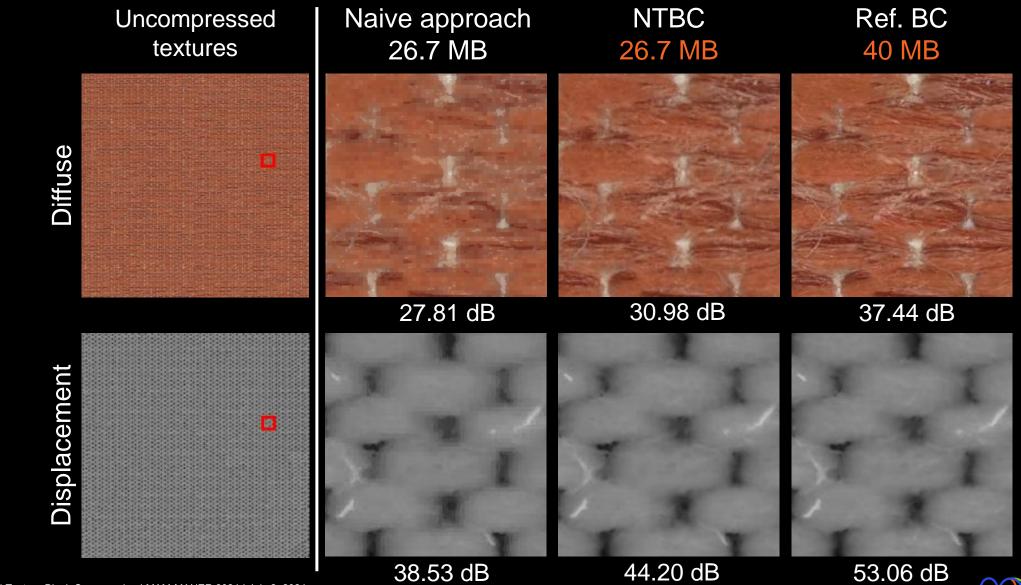
Uncompressed textures







Comparisons



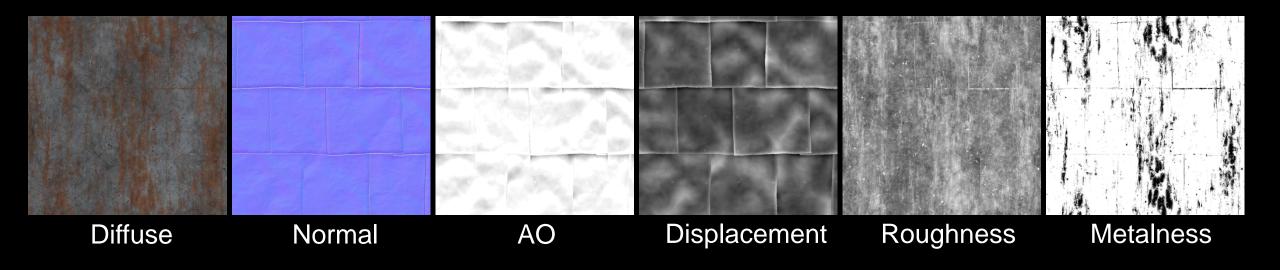
44.20 dB





Performance

- Evaluate training + inference performance with a single AMD Radeon[™] RX 7900 XT GPU
- Use "MetalPlates013" material
 - 2 RGB textures
 - 4 single-channel textures



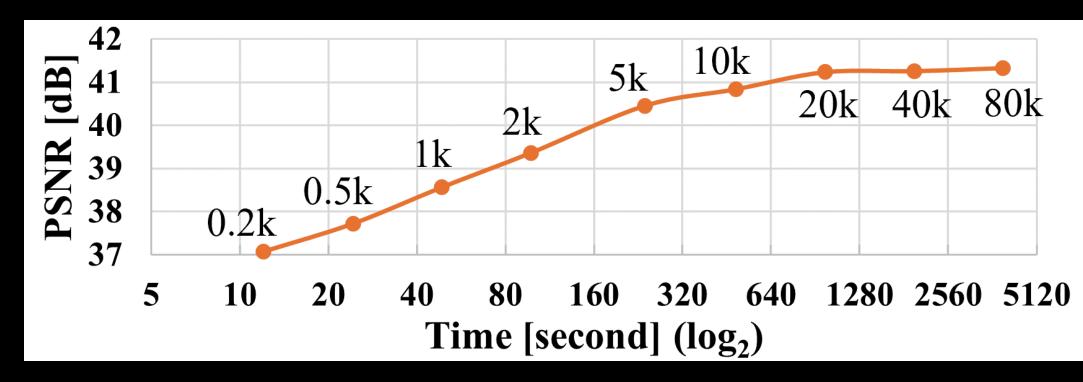


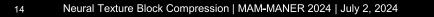




Performance - Training

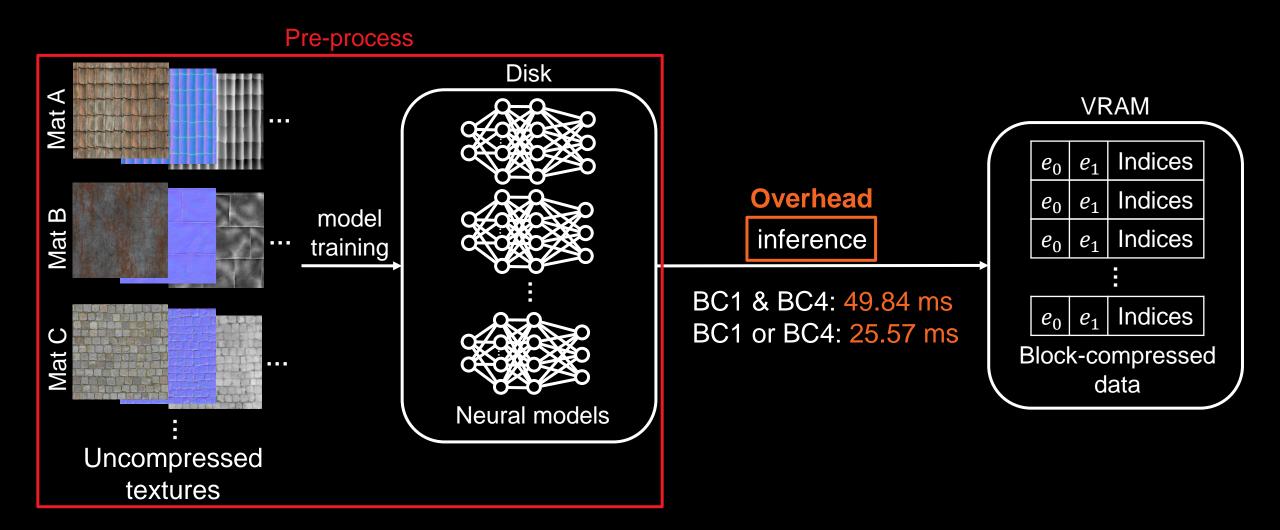
- NTBC training time over PSNR averaged over all the textures in the material
- 20k iterations gives saturated results in 16 minutes
- 2k iterations provides reasonable quality just in 100 seconds







Performance - Inference

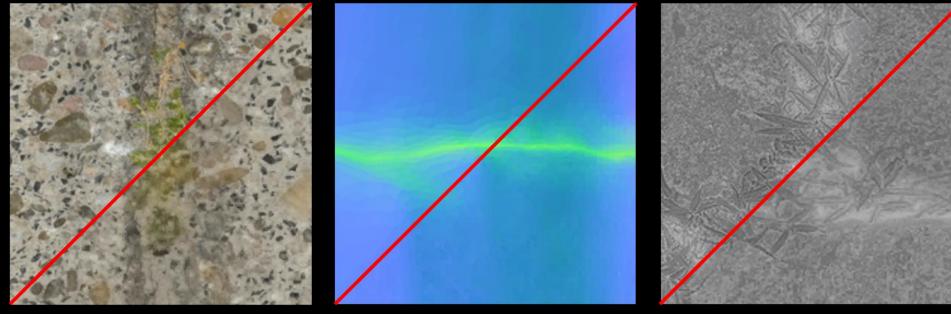




Limitations

R

- Color degradation
 - Happens if the texture contains high-frequency details both in luminance and colors
- Loss of detailed content
 - Block artifacts and blurred details
 - NTBC uses lower-resolution grids than the original texture, which could cause the errors



Color degradation

Block artifacts

Blurred detail





Future Work

- Input encoding specialized for textures
 - Use frequency information to handle high-frequency details efficiently
- Extension to more complex formats such as BC6H and BC7
 - BC1 and BC4 are simple but only show limited-quality results
- Mipmap support
- Evaluation in practical real-time applications



