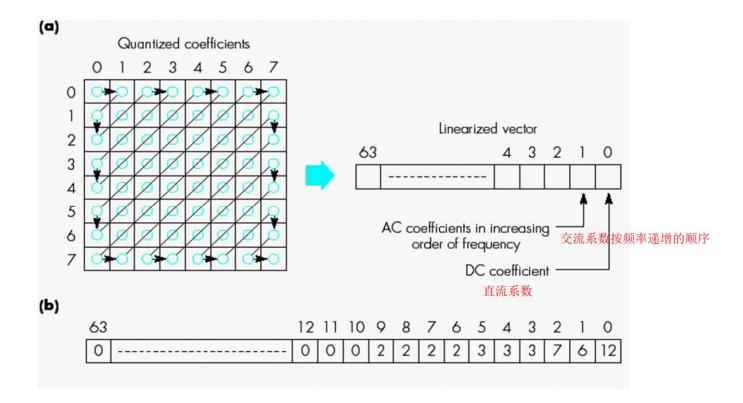
Zig-zag Scanning

锯齿形扫描



JPEG Encoder

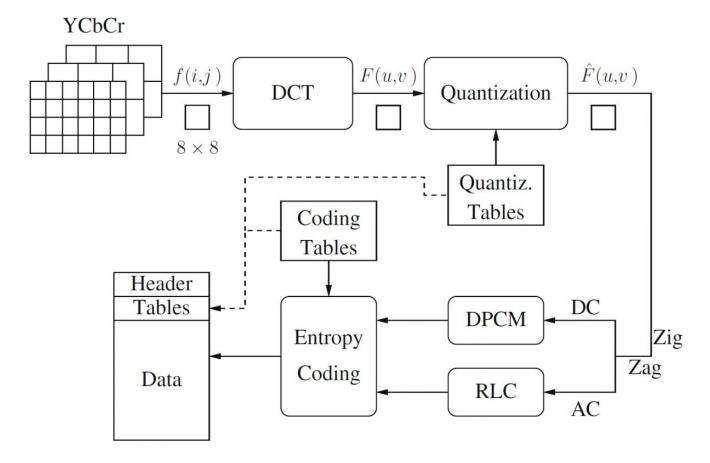
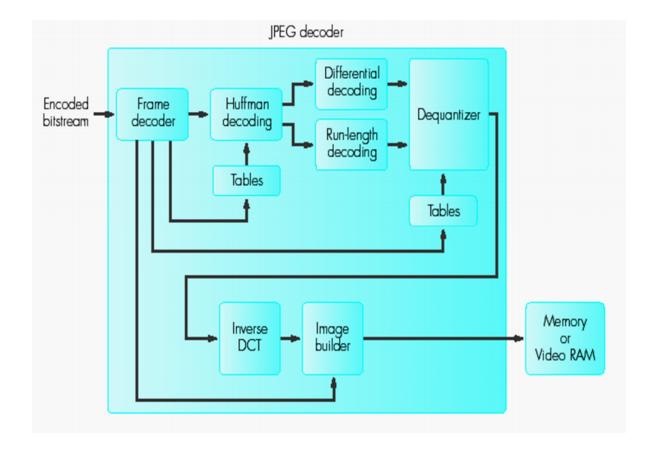


Fig. 9.1: Block diagram for JPEG encoder.

Source: Ze-Nian Li, Mark S. Drew, Jiangchuan Liu, Fundamental of Multimedia, Springer 2021

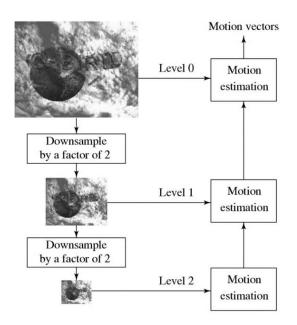
JPEG Decoder



Hierarchical Search

分层搜索; 阶层式搜寻

- A three-level hierarchical search, the original image is at Level 0.
- Images at Levels 1 and 2 are obtained by down-sampling from the previous levels by a factor of 2, and the initial search is conducted at Level 2.



A Three-level Hierarchical Search

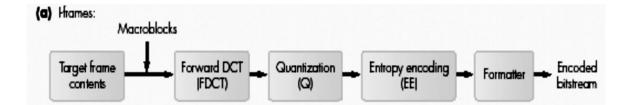
三级分层搜索,原始图像处于o级。

1级和2级的图像是通过对前两级进行2倍的降采 样获得的,初始搜索在2级进行。



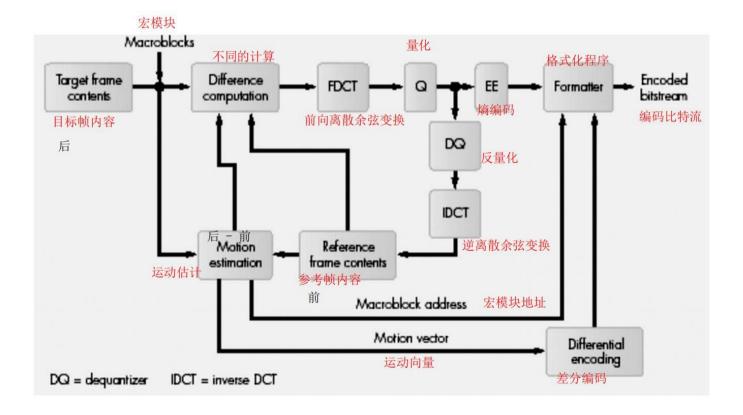
MPEG-1: I-Frame Encoding

I帧编码

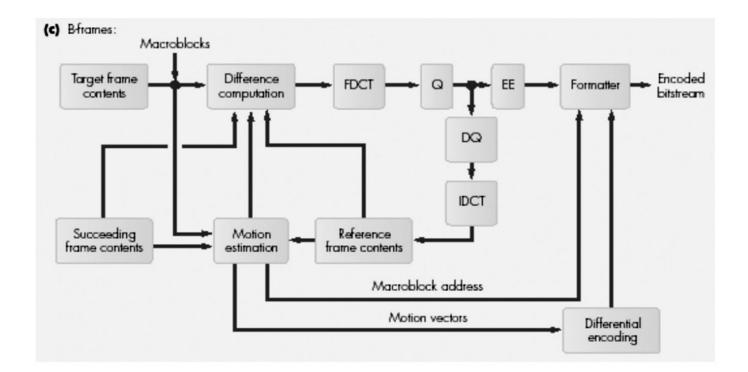


MPEG-1: P-Frame Encoding Flowchart

p帧编码流程图



MPEG-1: B-Frame Encoding Flowchart



Type of Scalabilities



可以传大的基础层

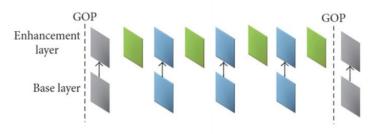
FIGURE 4: Spatial scalability.

图4:空间可扩展性。



基础层传垃圾

FIGURE 5: Quality scalability. 图5: 质量可伸缩性。



可以多传一倍的照片

FIGURE 6: Temporal scalability.

图6:时间可伸缩性。

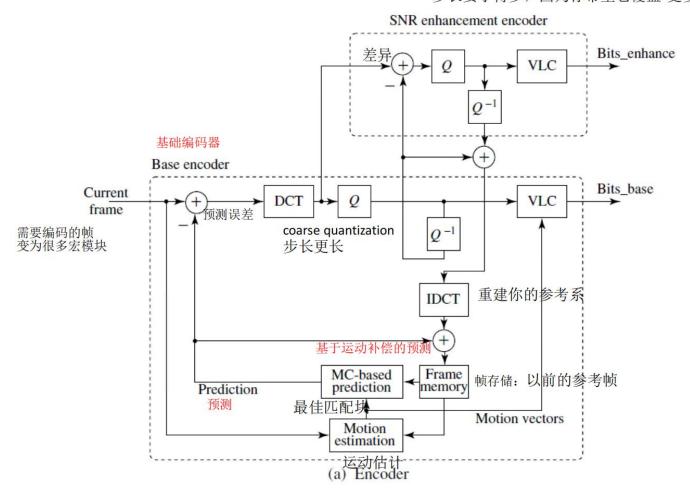
source: https://www.semanticscholar.org/paper/Efficient-Enhancement-for-Spatial-Scalable-Video-Khairy-Hamdy/d483c84609f393257dc2874aa86604a626242cd1



3.2.5.3.1.信噪比可扩展

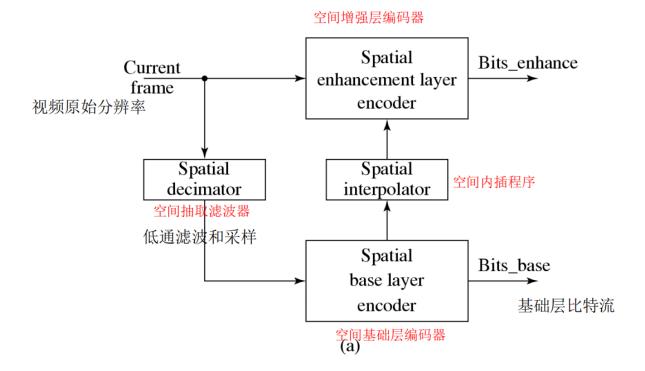
MPEG-2: SNR Scalability 2:13:41

步长要小得多,因为你希望它覆盖 更多细节

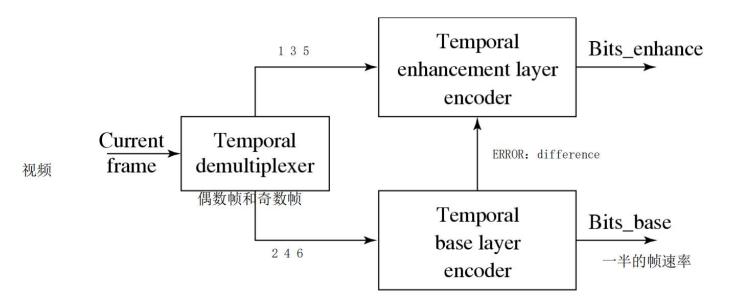


MPEG-2: Spatial Scalability 2:13:41

MPEG-2:空间可扩展性



MPEG-2: Temporal Scalability



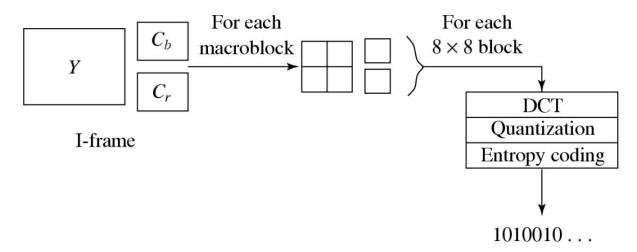
(a) Block Diagram

H.261: I-frame Coding

宏块包含4Y, 1Cb和1Cr的8 x 8像素块, 用于4:2:0色度子采样

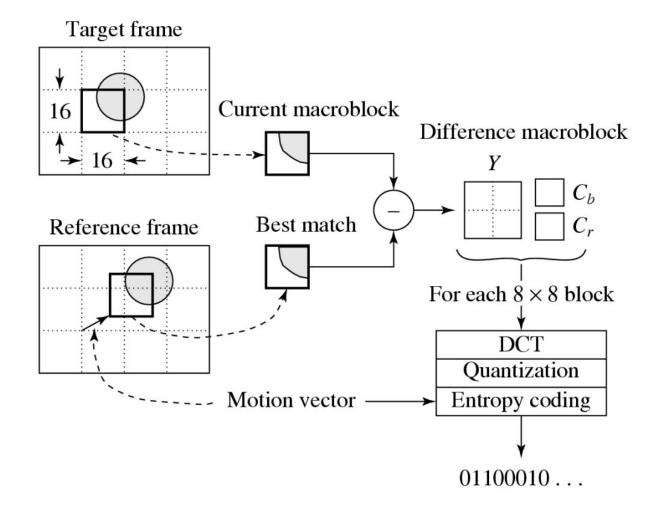
- A macroblock contains 4Y, 1Cb, and 1Cr of 8 x 8 pixel blocks for 4:2:0 chroma subsampling.
- Macroblocks are of size 16 x 16 pixels for the Y frame, and 8 x 8 for Cb and Cr frames. 对于Y帧,macroblock的大小为16 x 16像素,对于badcrres为8 x 8像素。
- For each 8 x 8 block, a DCT transform is applied, the DCT coefficients then go through quantization, zigzag scanning, and entropy coding.

对于每个8 x 8块,应用DCT变换,DCT系数然后经过量化,之字形扫描和熵编码。



I-frame Coding

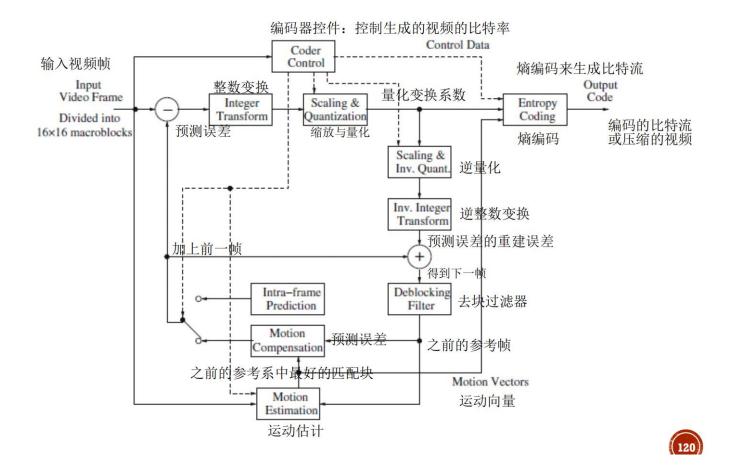
H.261: P-frame Coding



H.261 P-frame Coding Based on Motion Compensation

6427 week6 38:14 / 2:59:48

H.264: Encoder

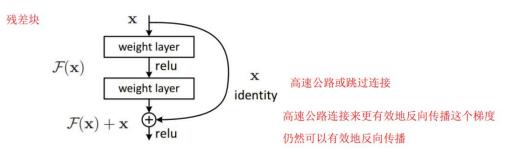


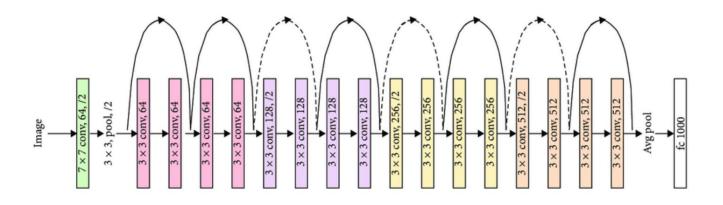
ResNet

消失梯度问题

继续向普通 CNN 添加许多不同的层,实际上性能在提高一段时间后就会开始下降

如果你将梯度堆叠在许多不同的层上, 是的,你的早期层无法有效更新

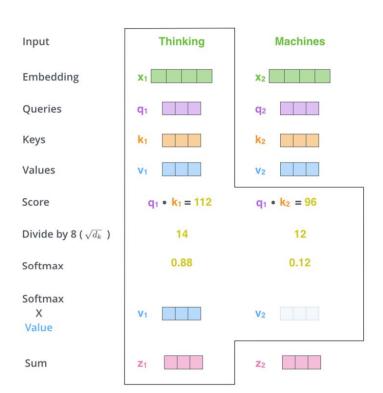




Scaled Dot-Product Attention

缩放的点积注意力

- Step 1: compute the correlation (dot product) between the query (q) and key (k) vectors.
- Step 2: correlation values from Step 1 are scaled and normalized using Softmax function.
- Step 3: multiplied output from Step 2 by corresponding value (v) vectors and sum them up.



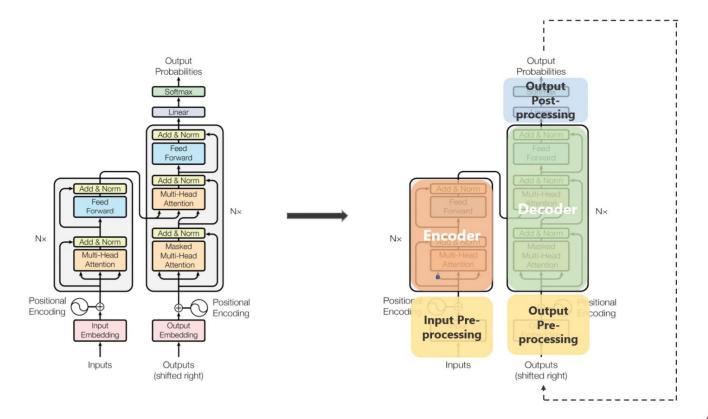
步骤1:计算查询(q)和键(k)向量之间的相关性(点积)。 步骤2:使用Softmax函数对步骤1的相关值进行缩放和归一化。

步骤3:将步骤2的输出与相应的值向量(v)相乘并求和。

$$A(Q,K,V) = softmax \Big(\frac{QK^T}{\sqrt{d_k}}\Big)V$$
 softmax $\Big(\frac{\mathbf{V}}{\sqrt{d_k}}\Big)$



Transformer Architecture (1) 变压器结构



Source: Ria Kulshrestha, Transformers. https://towardsdatascience.com/transformers-89034557de14

