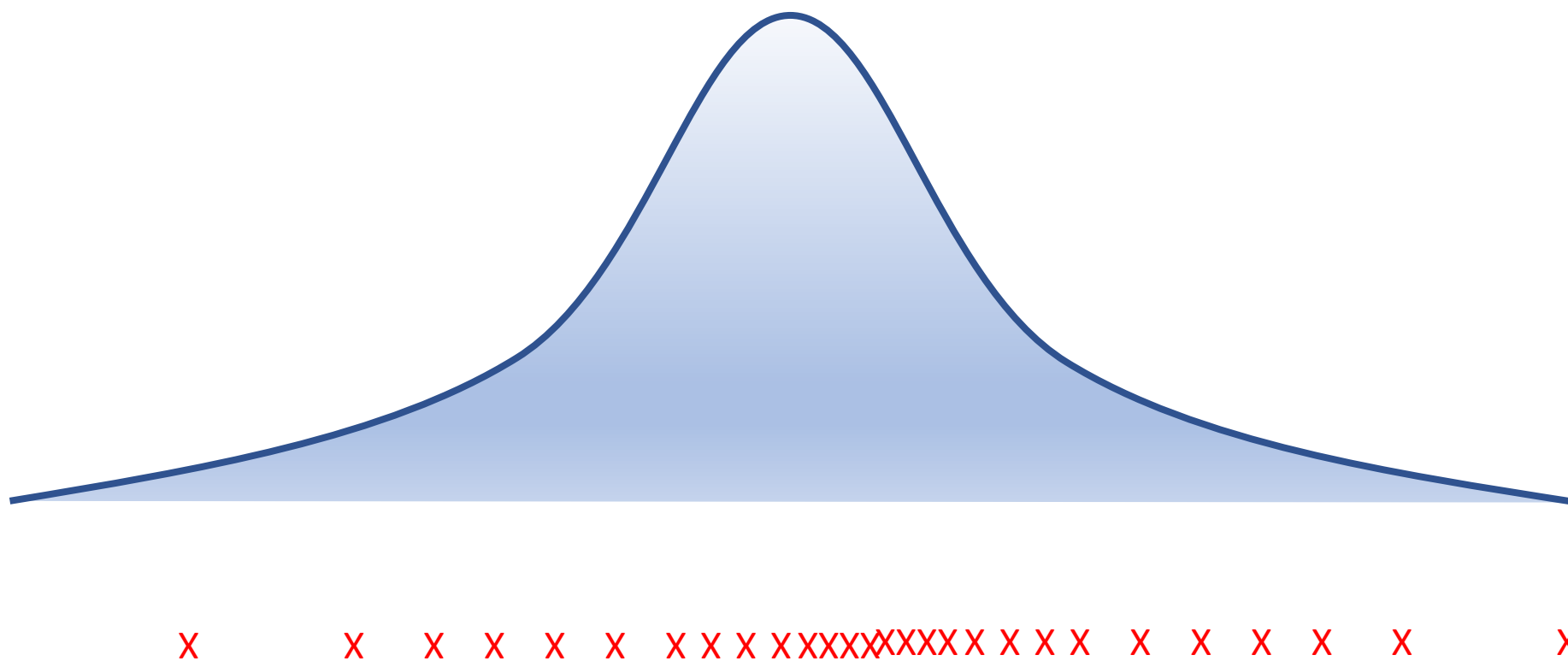


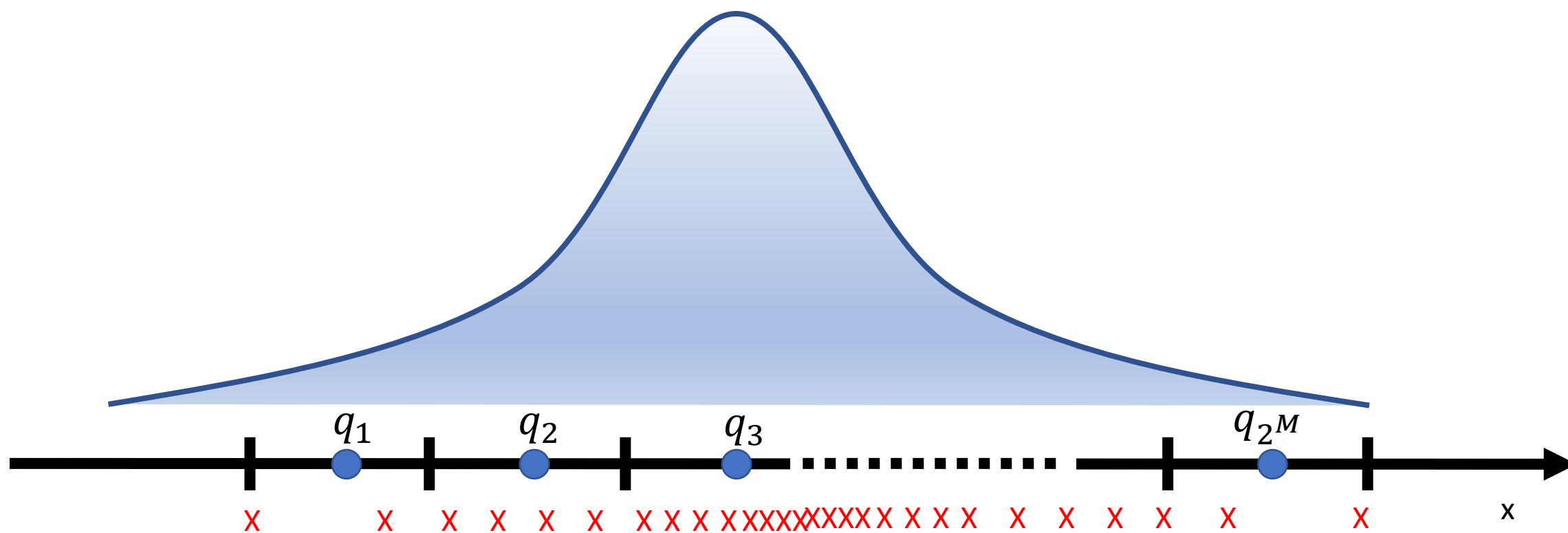
Post training 4-bit quantization of convolutional networks for rapid-deployment

Ron Banner, Yury Nahshan and Daniel Soudry

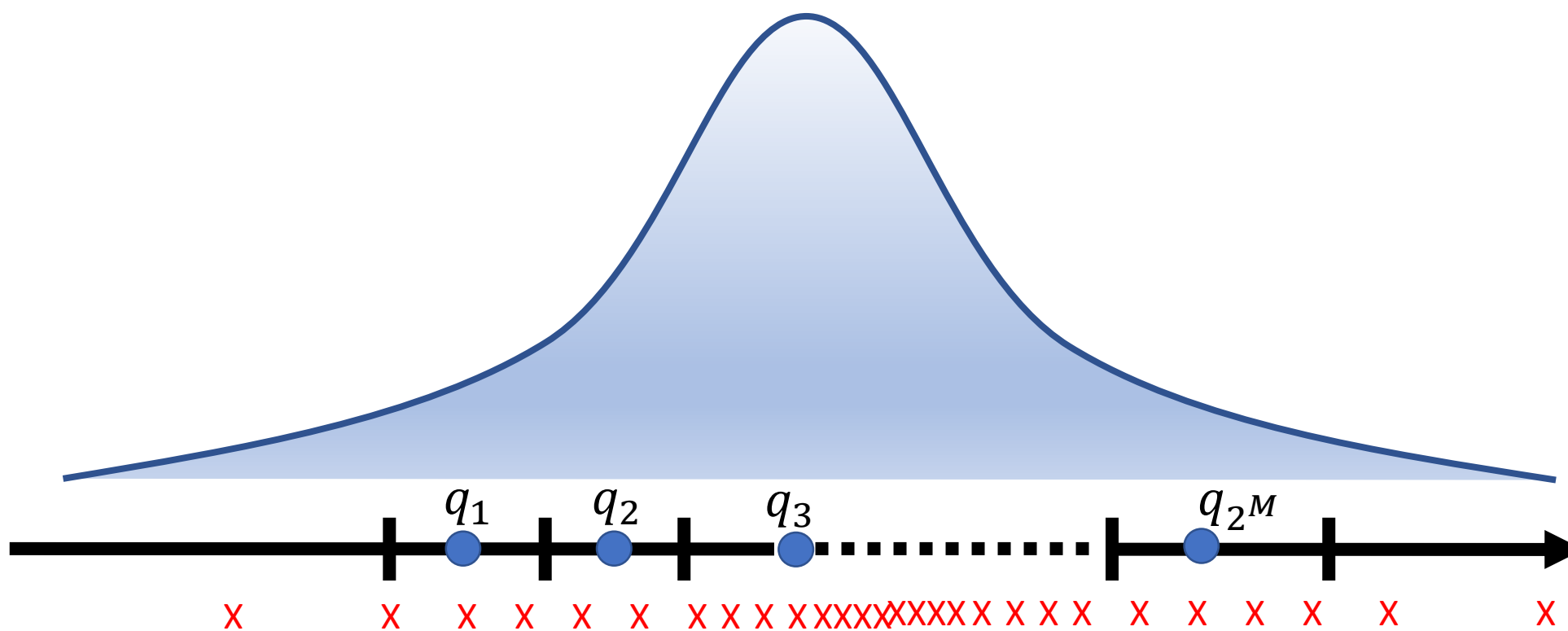
Pre-activation tensor



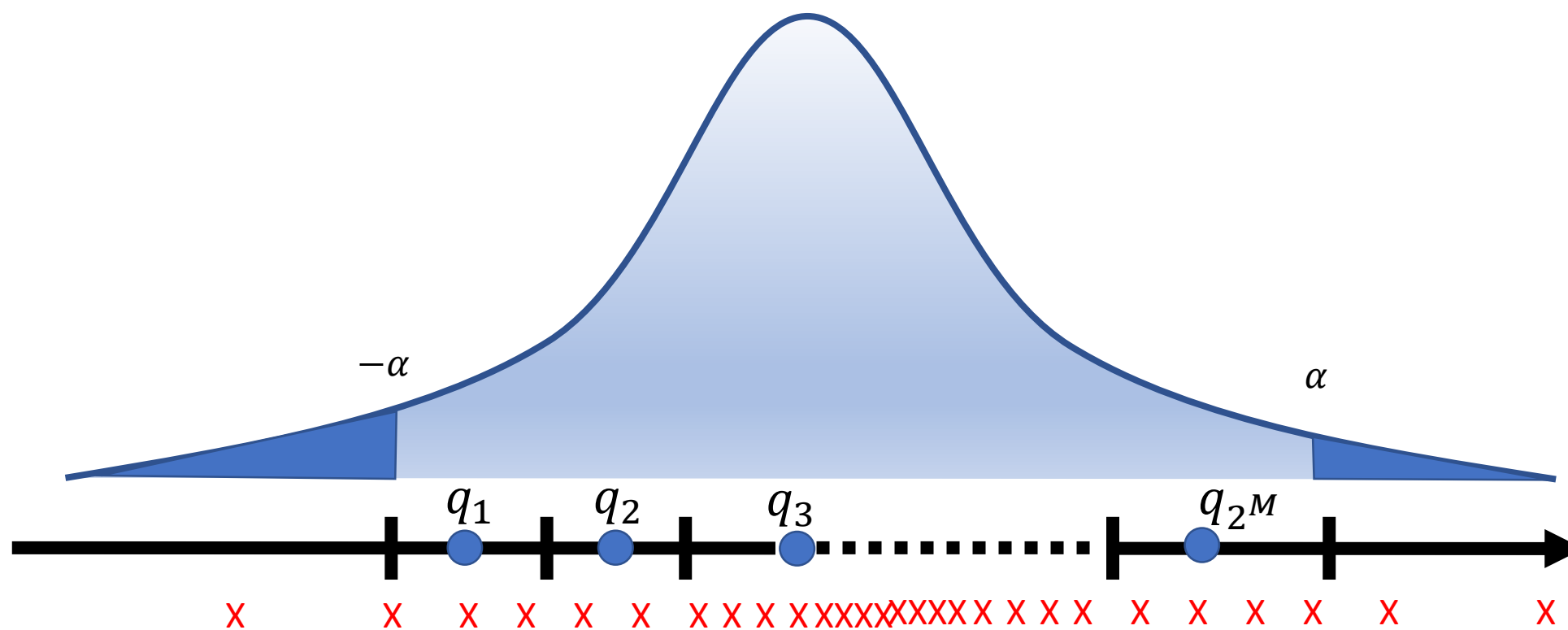
Pre-activation tensor

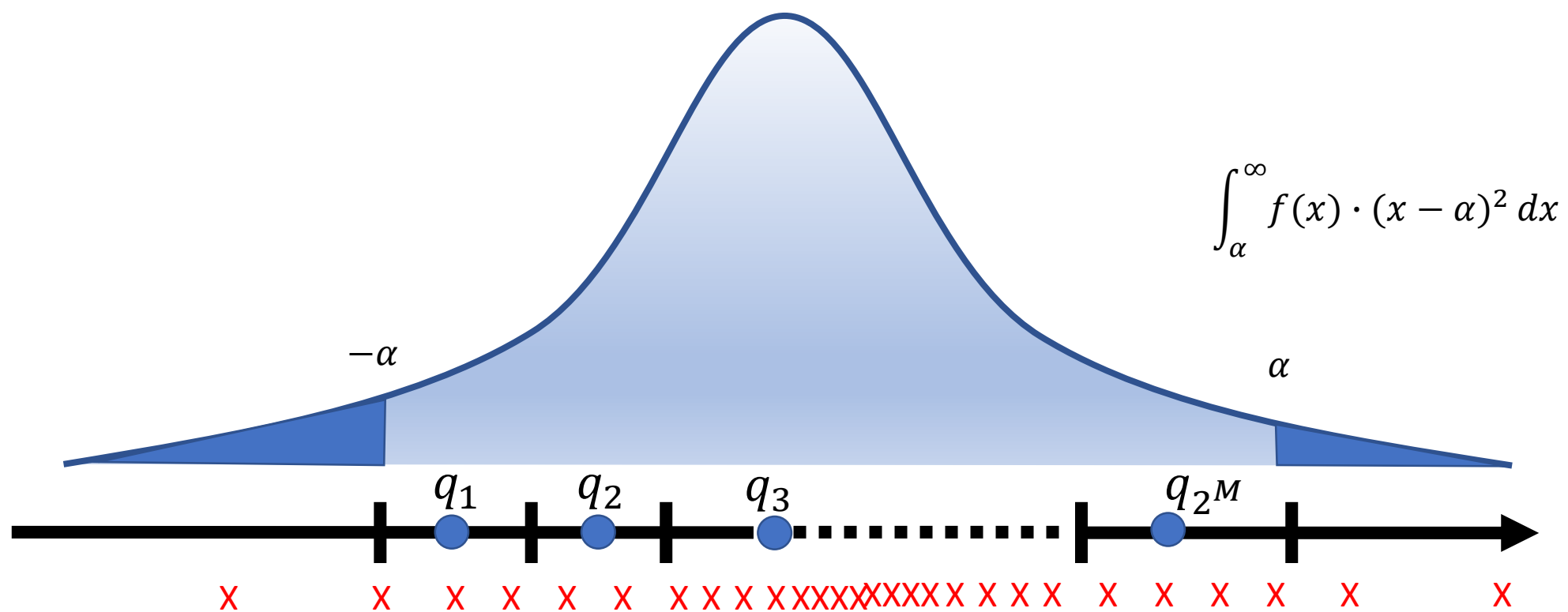


Pre-activation tensor

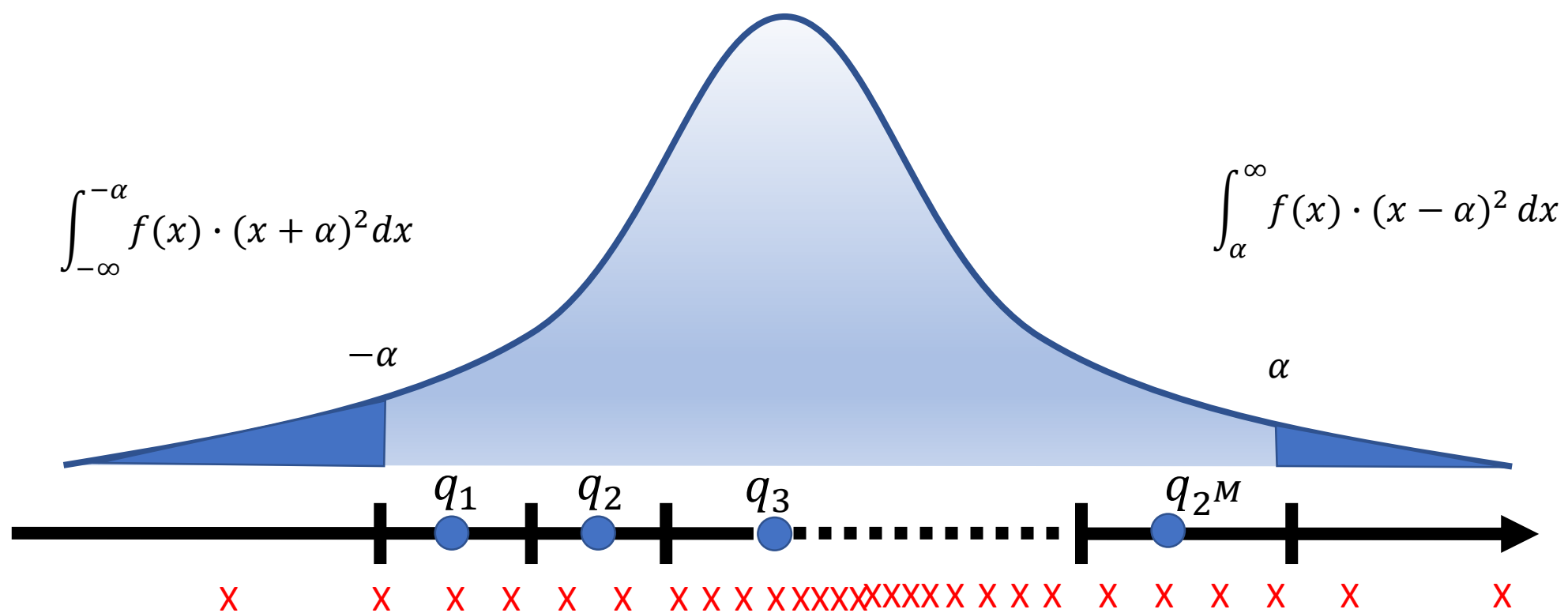


Pre-activation tensor

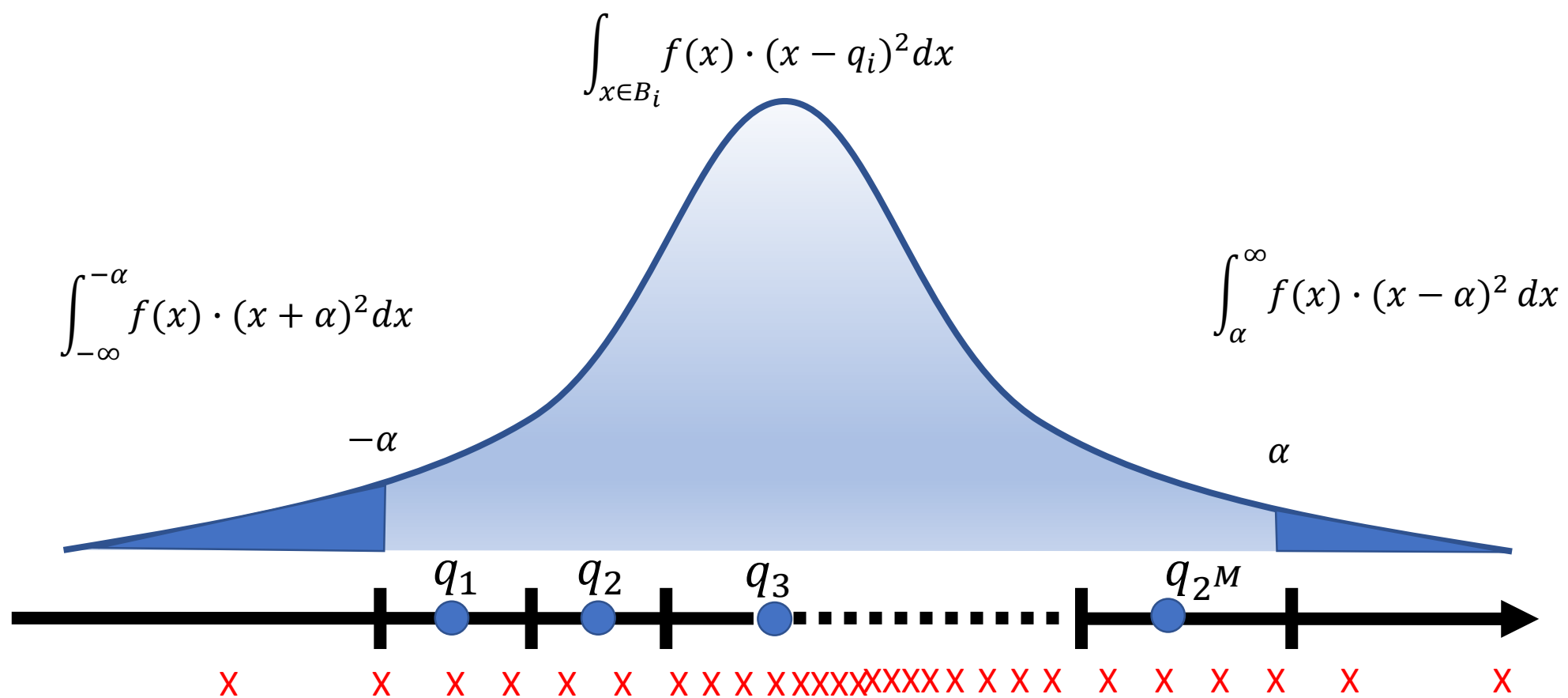




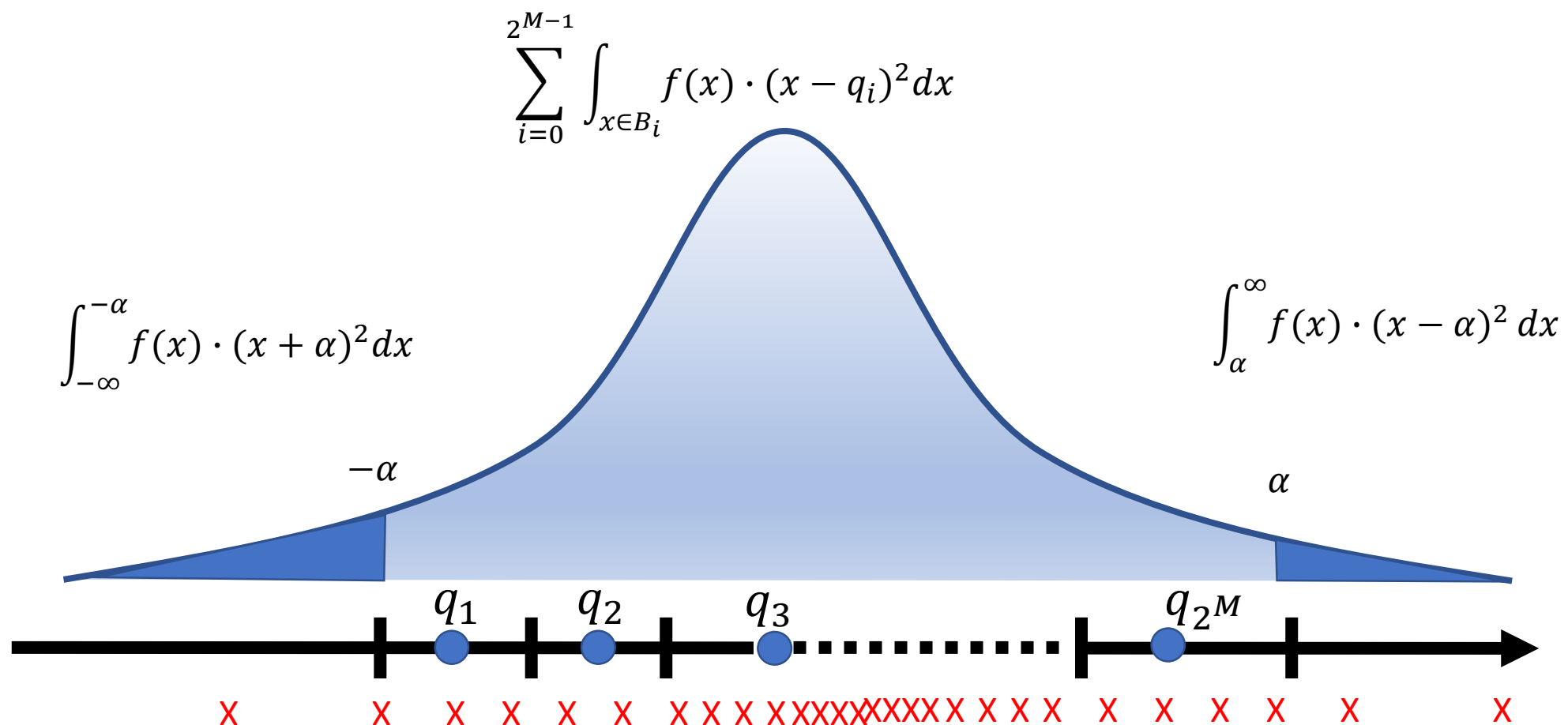
Pre-activation tensor



Pre-activation tensor



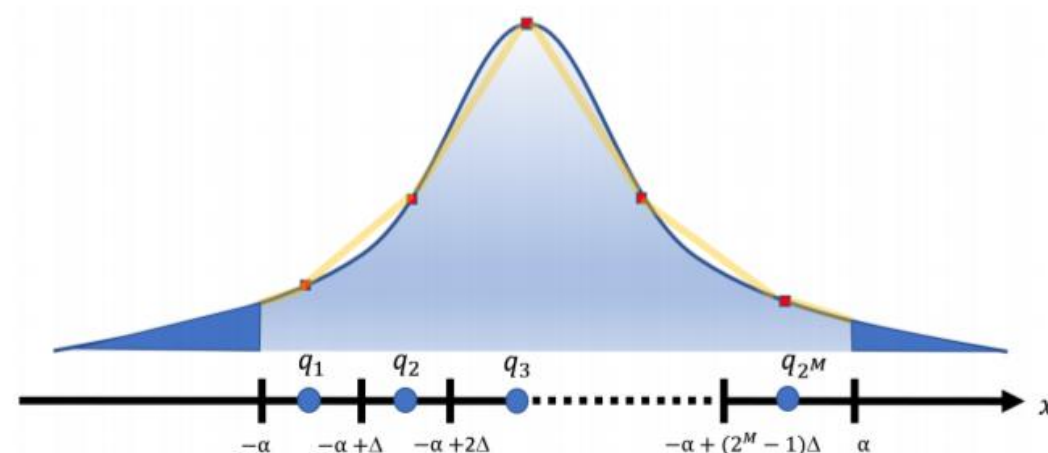
Pre-activation tensor



Optimal clipping value

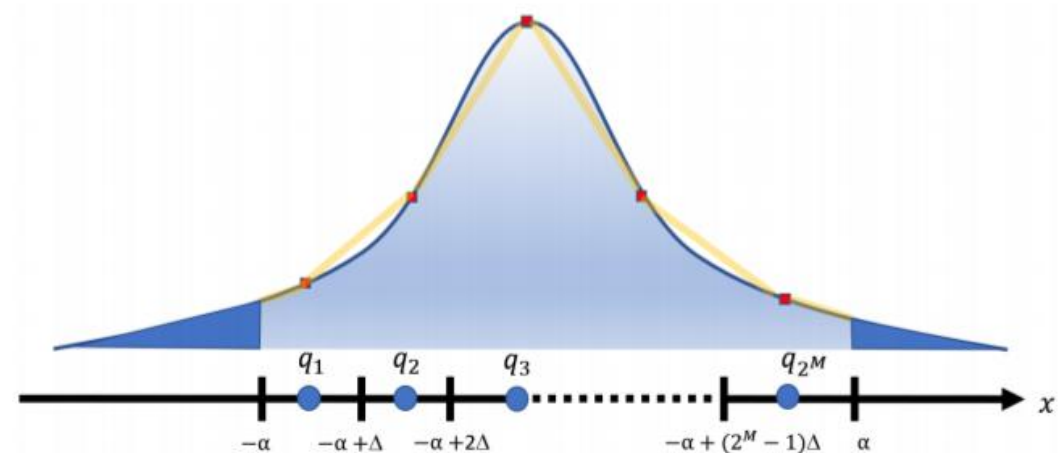


$$\begin{aligned} E[(X - Q(X))^2] &= \\ &= \int_{-\infty}^{-\alpha} f(x) \cdot (x + \alpha)^2 dx + \\ &+ \sum_{i=0}^{2^M-1} \int_{-\alpha+i\Delta}^{-\alpha+(i+1)\Delta} f(x) \cdot (x - q_i)^2 dx + \\ &+ \int_{\alpha}^{\infty} f(x) \cdot (x - \alpha)^2 dx \end{aligned}$$



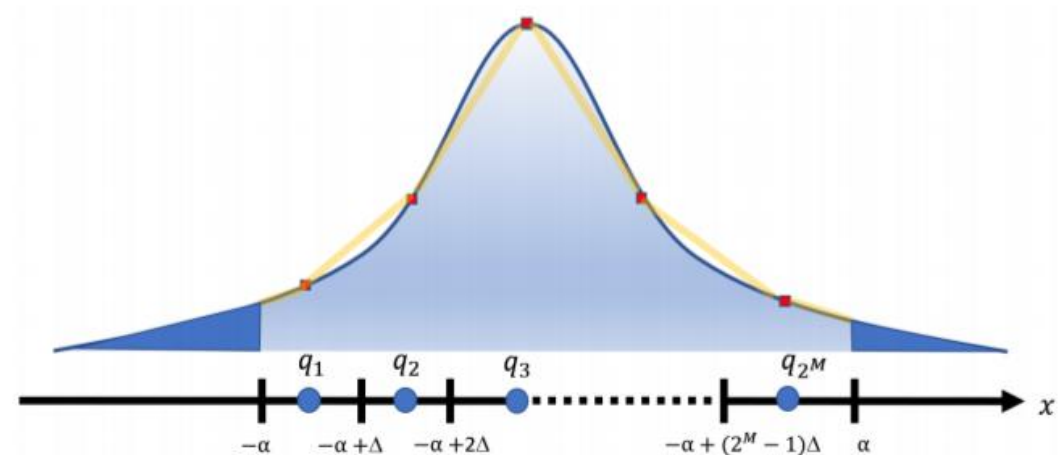
Optimal clipping value

$$\begin{aligned} E[(X - Q(X))^2] &\approx \\ &\approx (\alpha^2 + \sigma^2) \cdot \left[1 - \operatorname{erf} \left(\frac{\alpha}{\sqrt{2}\sigma} \right) \right] + \\ &+ \frac{\alpha^2}{3 \cdot 2^{2M}} - \frac{\sqrt{2}\alpha \cdot \sigma \cdot e^{-\frac{\alpha^2}{2 \cdot \sigma^2}}}{\sqrt{\pi}} \end{aligned}$$



Optimal clipping value

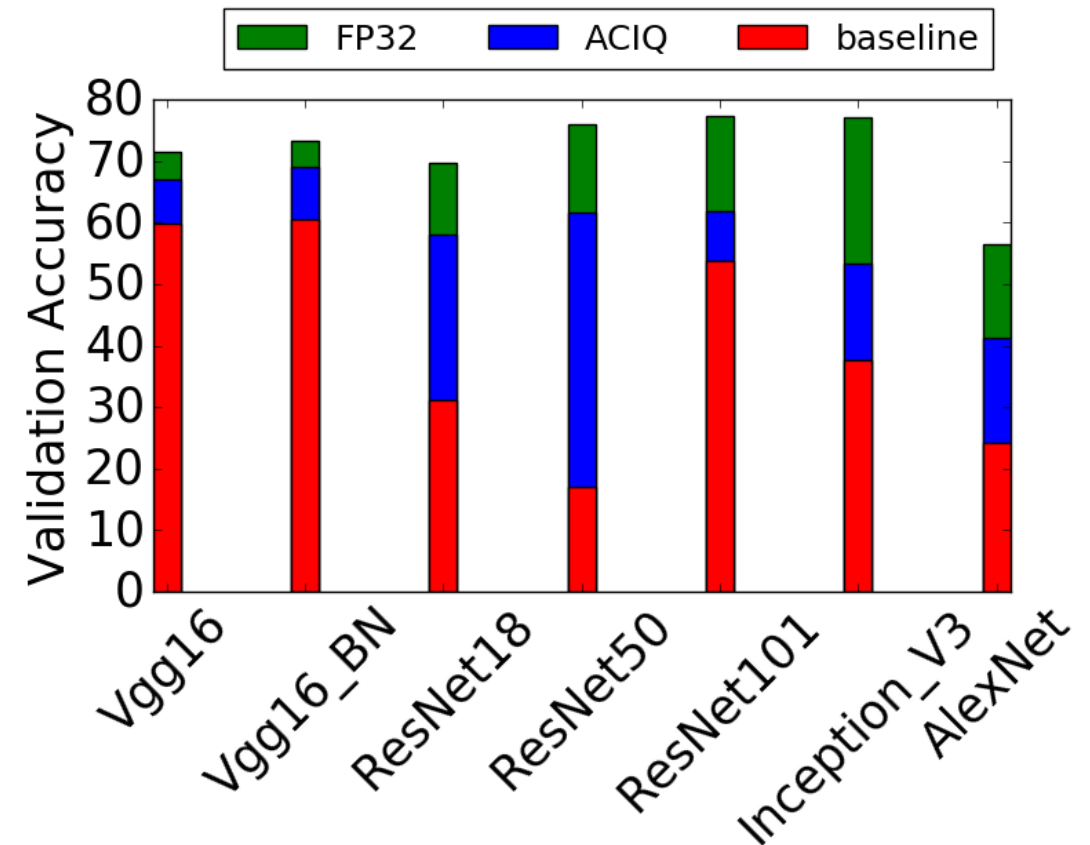
$$\begin{aligned}\frac{\partial E[(X - Q(X))^2]}{\partial \alpha} &= \\ &= \alpha \left[1 - \operatorname{erf} \left(\frac{\alpha}{\sqrt{2}\sigma} \right) \right] - \frac{\sigma^2 e^{-\frac{\alpha^2}{2\sigma^2}}}{\sqrt{2\pi}\sigma} - \frac{\sigma e^{-\frac{\alpha^2}{2\sigma^2}}}{\sqrt{2\pi}} + \\ &+ \frac{2\alpha}{3 \cdot 2^{2M}} = 0\end{aligned}$$



4-bit quantization

Model	Reference (FP32)	Ours (Optimal Clip)	GEMMLOWP (Max/Min)
VGG16	71.59%	70.1%	68.8%
VGG16-BN	73.36%	72.0%	70.6%
ResNet18	69.75%	66.6%	61.5%
ResNet50	76.1%	71.8%	68.3%
ResNet101	77.3%	72.6%	66.5%
Inception V3	77.2%	72.7%	70.9%

3-bit quantization



Per-Channel Bit allocation

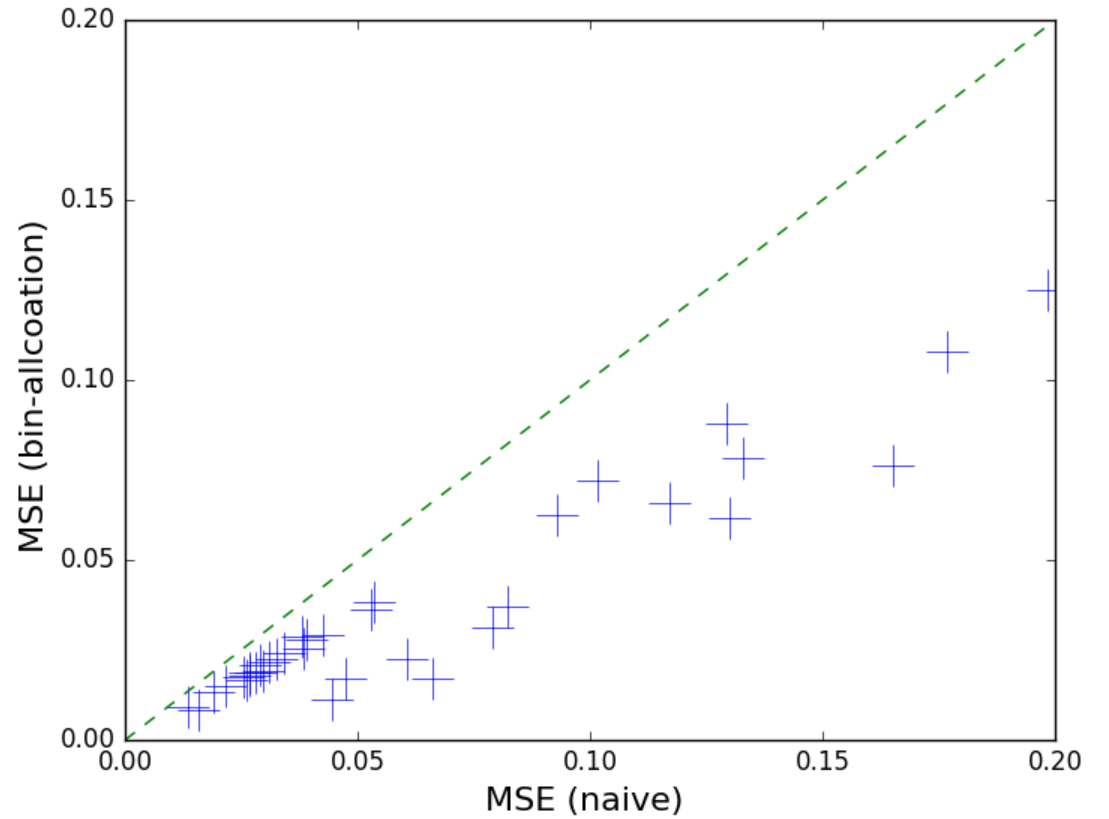
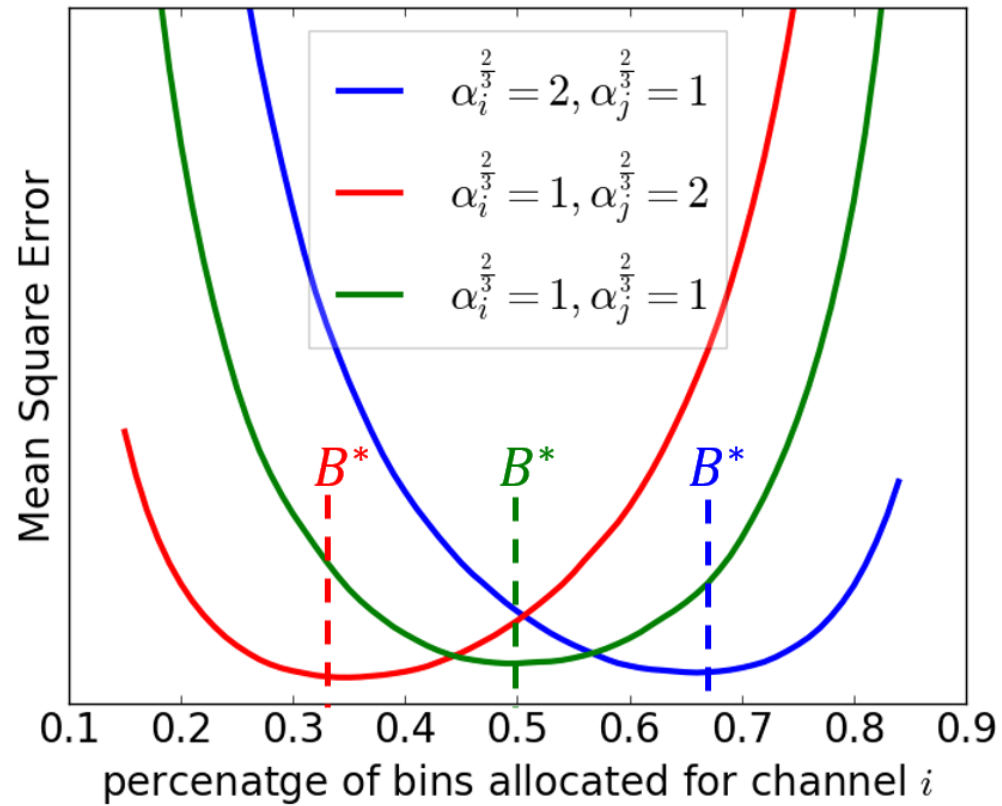
Problem:

Given a maximum bit-budget B , how many bits M_i should we allocate to each channel in order to minimize the layer mean-square-error ?

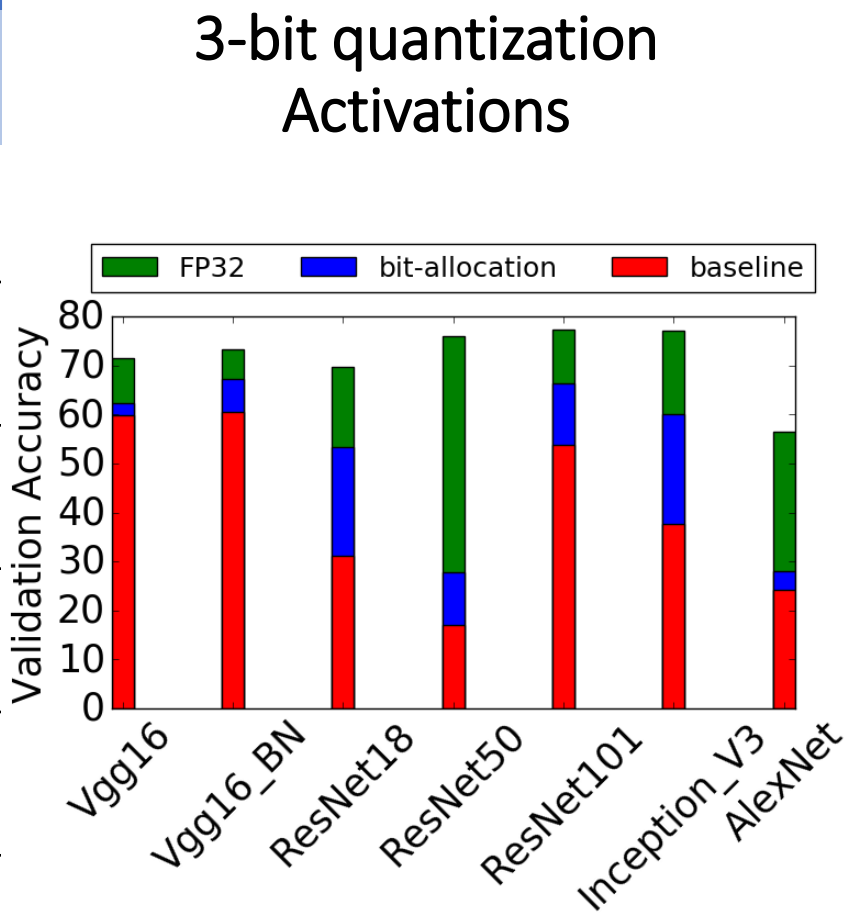
Answer:

$$M_i = \left\lfloor \log_2 \left(\frac{\alpha_i^{\frac{2}{3}}}{\sum_i \alpha_i^{\frac{2}{3}}} \cdot B \right) \right\rfloor$$

Per-Channel Bit allocation



4-bit quantization Activation				4-bit quantization Weights			
Model	Reference (FP32)	Ours (Bit-allocation)	Naive	Model	Reference (FP32)	Ours (Bit-allocation)	Naive
VGG16	71.59%	69.7%	68.8%	VGG16	71.59%	71.0%	70.5%
VGG16-BN	73.36%	72.6%	70.6%	VGG16-BN	73.36%	71.9%	68.5%
ResNet18	69.75%	65.0%	61.5%	ResNet18	69.75%	66.7%	59.7%
ResNet50	76.1%	71.3%	68.3%	ResNet50	76.1%	75.0%	72.5%
ResNet101	77.3%	70.8%	66.5%	ResNet101	77.3%	76.4%	74.6%
Inception V3	77.2%	74.3%	70.9%	Inception V3	77.2%	61.4%	38.4%



Bias-Correction

Problem:

We observe an inherent bias in the mean of the weight values following their quantization

Solution:

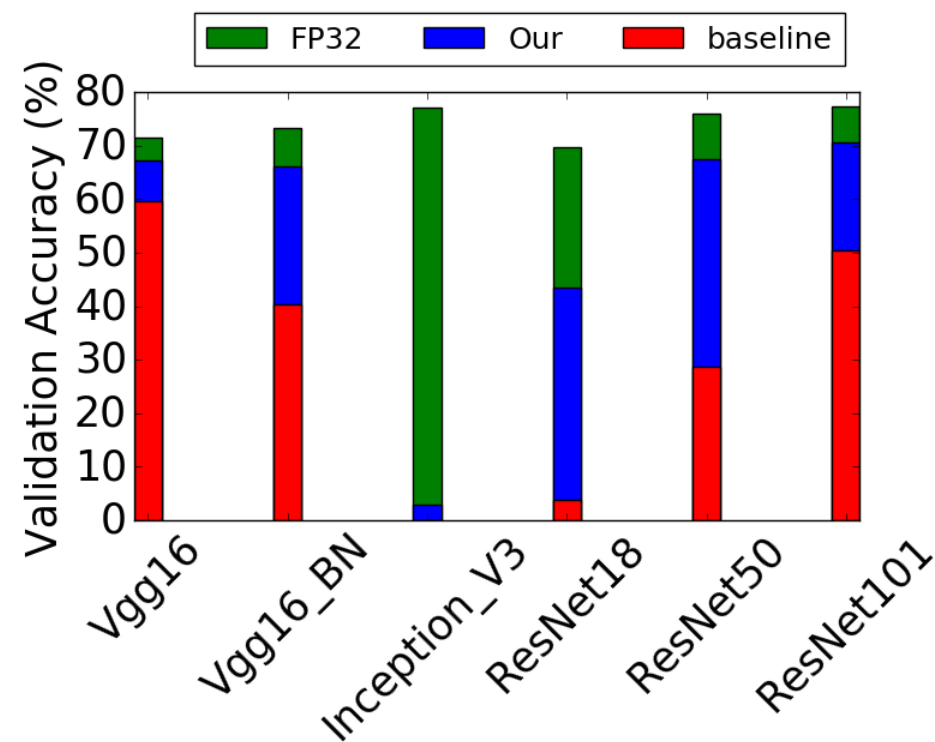
$$\mu_c = \mathbb{E}(W_c) - \mathbb{E}(W_c^q)$$
$$\xi_c = \frac{\|W_c - \mathbb{E}(W_c)\|_2}{\|W_c^q - \mathbb{E}(W_c^q)\|_2}$$

$$w \longleftarrow \xi_c (w + \mu_c), \quad \forall w \in W_c^q$$

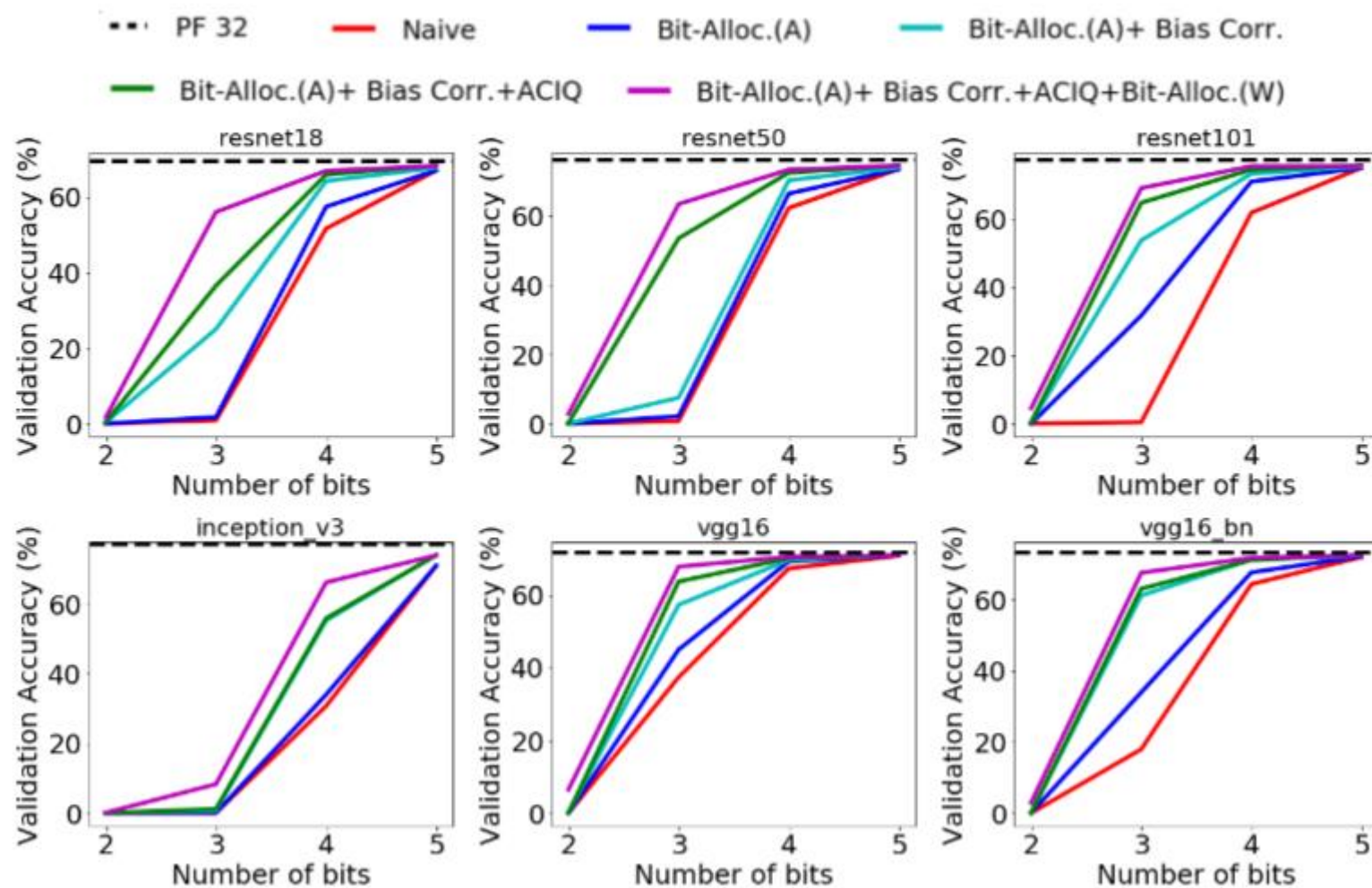
4-bit quantization Weights

Model	Reference (FP32)	Ours (Bias-correction)	Naive
VGG16	71.59%	71.0%	70.5%
VGG16-BN	73.36%	71.7%	68.5%
ResNet18	69.75%	67.4%	59.7%
ResNet50	76.1%	74.8%	72.5%
ResNet101	77.3%	76.3%	74.6%
Inception V3	77.2%	59.5%	38.4%

3-bit quantization



All method combined



All methods combined

- how much can we get without hurting 8-bit baseline?

No re-training	
Model	Avg number of bits per value (weights and activations)
VGG16	5.3 bits
VGG16-BN	5.4 bits
ResNet18	5.4 bits
ResNet50	5.7 bits
ResNet101	5.8 bits
Inception V3	6.1 bits

All methods combined

- how much can we get without hurting 8-bit baseline?
- Unless we support variable-length coding, **we need to support different bit-width per channel**

No re-training	
Model	Avg number of bits per value (weights and activations)
VGG16	5.3 bits
VGG16-BN	5.4 bits
ResNet18	5.4 bits
ResNet50	5.7 bits
ResNet101	5.8 bits
Inception V3	6.1 bits

All methods combined

- how much can we get without hurting 8-bit baseline?
- Unless we support variable-length coding, **we need to support different bit-width per channel**
- Variable-length coding can further improve these results ...

No re-training	
Model	Avg number of bits per value (weights and activations)
VGG16	5.3 bits
VGG16-BN	5.4 bits
ResNet18	5.4 bits
ResNet50	5.7 bits
ResNet101	5.8 bits
Inception V3	6.1 bits

How much can we get without hurting 8-bit baseline?

No re-training		
Model	Avg number of bits per value (without VLC)	Avg number of bits per value (with VLC)
VGG16	5.3 bits	2.2 bits
VGG16-BN	5.4 bits	2.4 bits
ResNet18	5.4 bits	3.7 bits
ResNet50	5.7 bits	4.1 bits
ResNet101	5.8 bits	4.4 bits
Inception V3	6.1 bits	3.1 bits

End