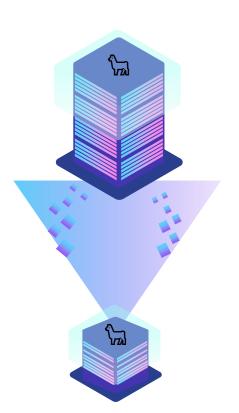
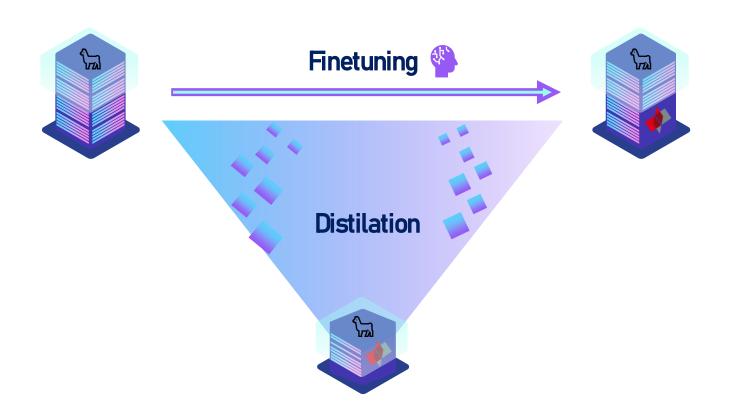
# Distilling Llama 2

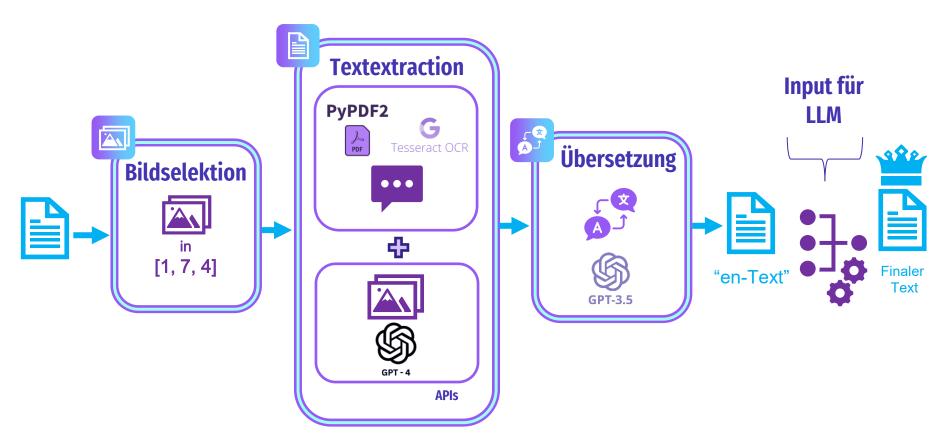
Vom Lehrer zum Schüler



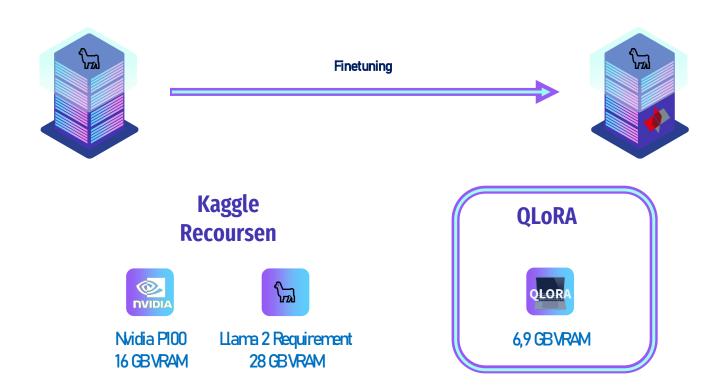
## **Zielsetzung**



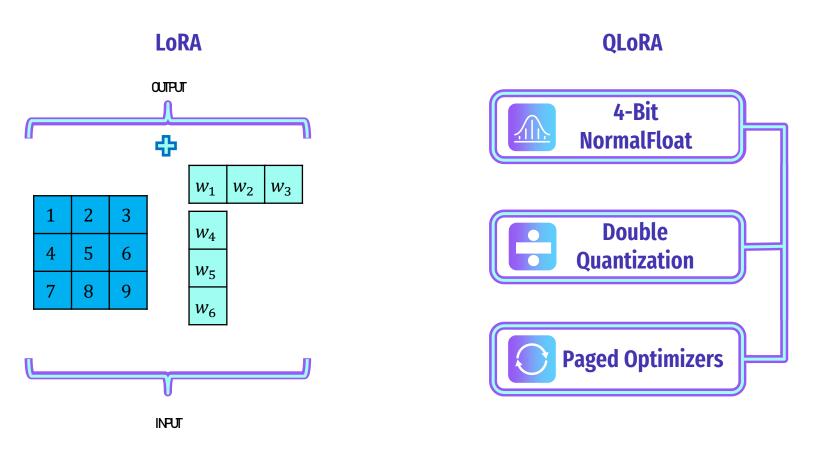
## **Data Preprocessing**



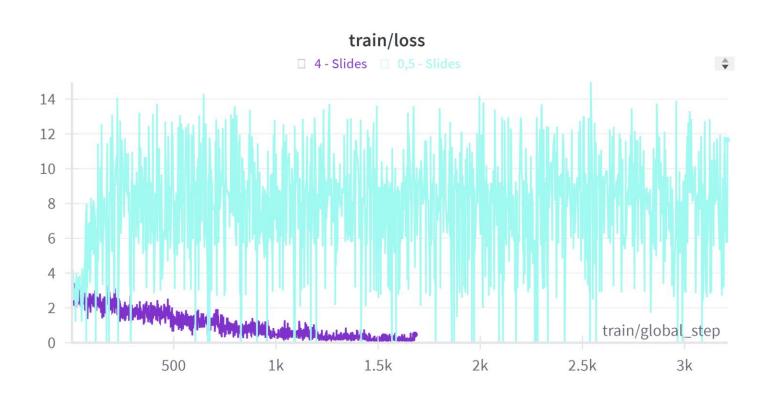
## **Finetuning Llama 2**



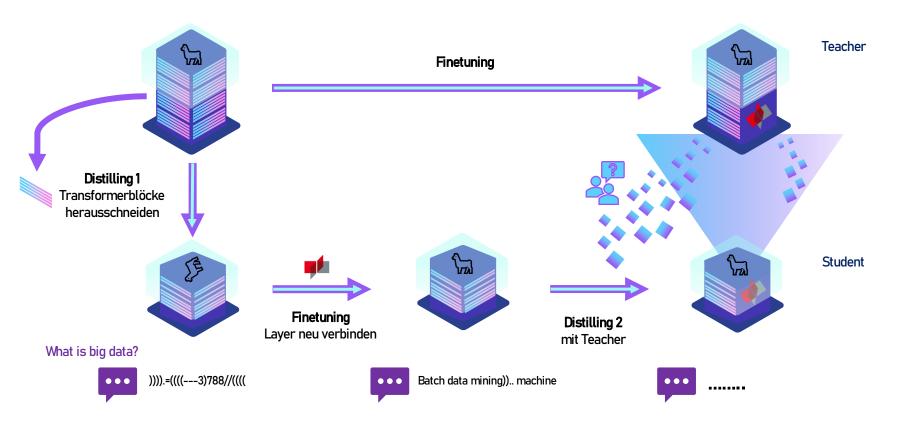
## Finetuning Llama 2



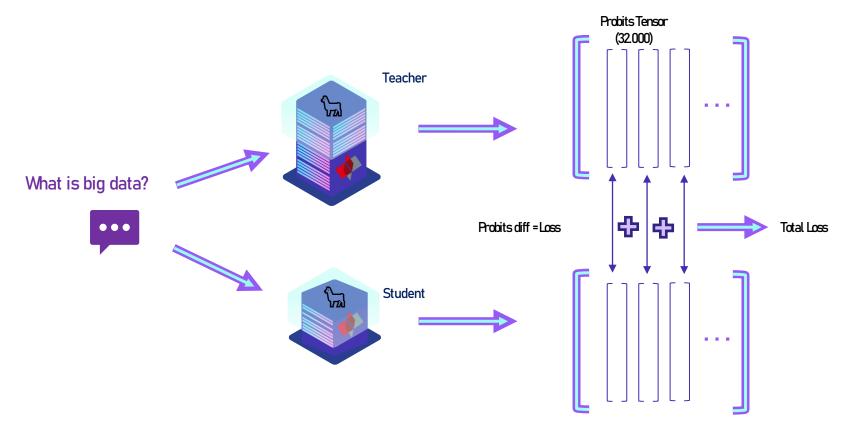
## **Finetuning Llama 2**

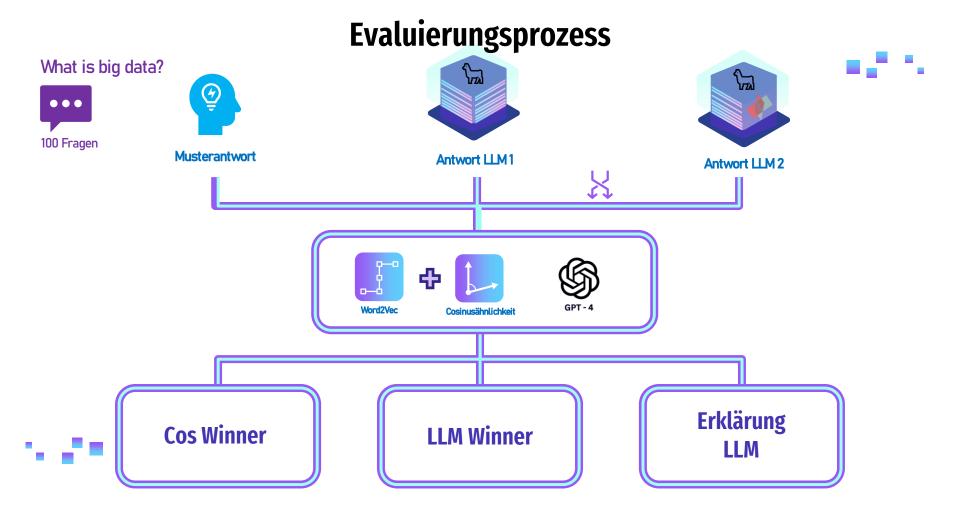


## **Distilling Llama 2**

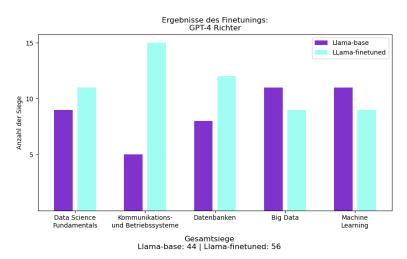


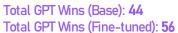
## **Distilling Llama 2**

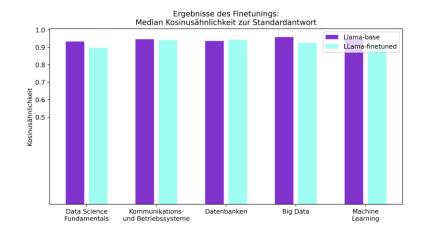




## **Ergebnisse Finetuning**







Total Similarity Wins (Base): **66**Total Similarity Wins (Fine-tuned): **34** 





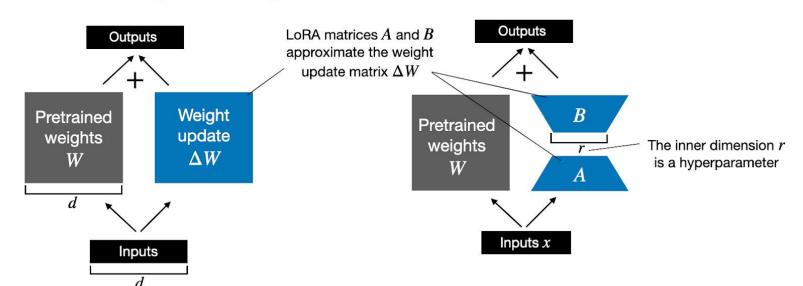


# Backup Folien:

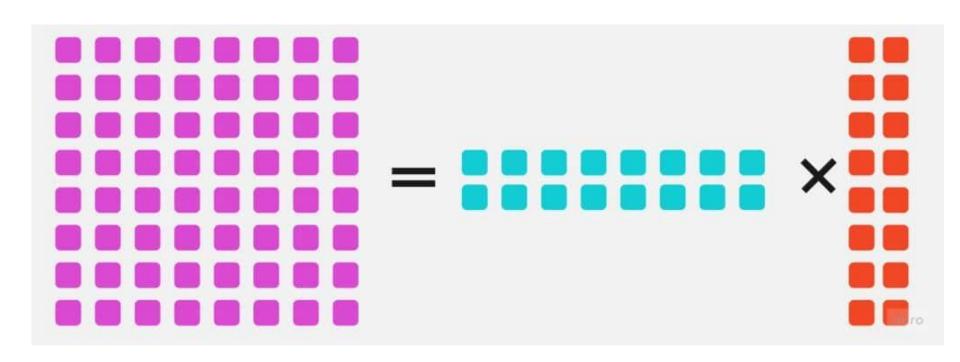
### **LoRA**

#### Weight update in regular finetuning

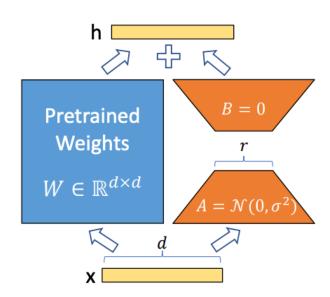
#### Weight update in LoRA

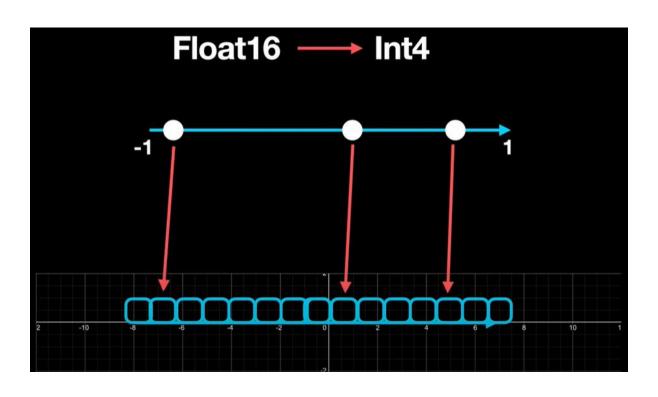


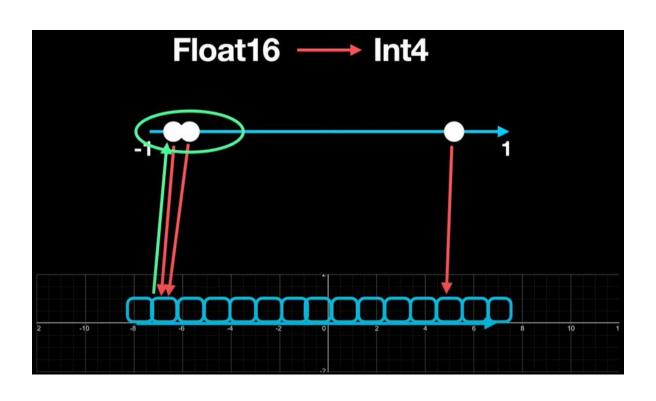
### **LoRA**

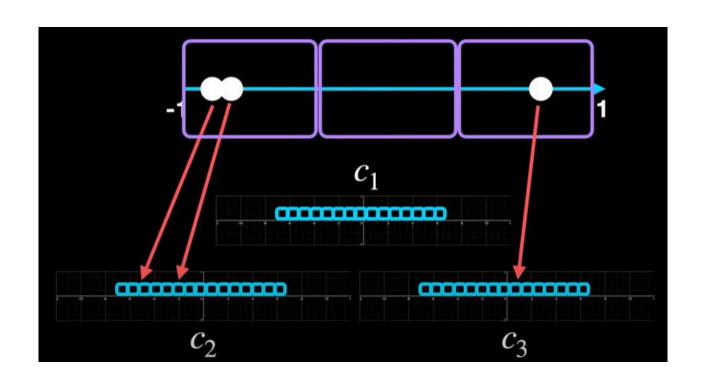


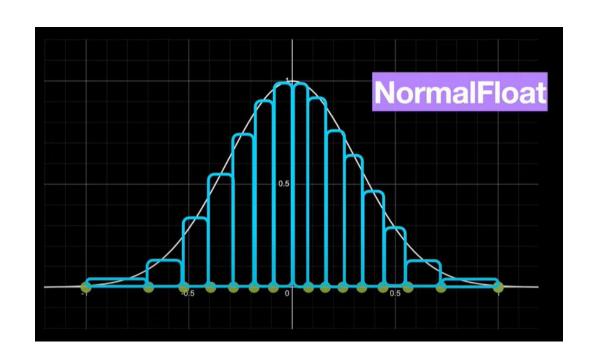
## **LoRA**











## **QLoRA: Double Quantization**

**Double Quantization** We introduce *Double Quantization* (DQ), the process of quantizing the quantization constants for additional memory savings. While a small blocksize is required for precise 4-bit quantization [13], it also has a considerable memory overhead. For example, using 32-bit constants and a blocksize of 64 for  $\mathbf{W}$ , quantization constants add 32/64 = 0.5 bits per parameter on average. Double Quantization helps reduce the memory footprint of quantization constants.

## **QLoRA: Paged Optimizers**



# **Ergebnisse Destillation**