

Instructions: Select the best answer(s). Multi-select where noted.

- Q1. (Multi-select)** Which are **true** about pre-training in LLMs?
- A. Maximizes log-likelihood of a token sequence given its prefix
 - B. Results in instruction-following behavior directly
 - C. Produces base models for downstream fine-tuning
 - D. Optimizes human preference rankings
- Q2.** Instruction tuning is most effective when:
- A. The model classifies instruction types
 - B. Fine-tuned on diverse input-output instruction pairs
 - C. The model is distilled first
 - D. Loss is replaced with an RL reward signal
- Q3. (Multi-select)** In RLHF, the reward model is trained by:
- A. Cross-entropy on token prediction
 - B. Learning scalar rewards from preference comparisons
 - C. Binary classification between preferred and dispreferred responses
 - D. RL training from scratch
- Q4.** PPO is used **after**:
- A. Pretraining and before instruction tuning
 - B. Reward model training from preferences
 - C. Direct preference optimization
 - D. Few-shot prompting
- Q5.** PPO uses this to avoid destructive updates:
- A. KL regularization
 - B. Entropy bonus
 - C. Value baseline
 - D. Clipped probability ratio
- Q6.** DPO avoids reward modeling by:
- A. Actor-critic methods
 - B. Contrastive loss on preference pairs
 - C. Sampling rewards until saturation
 - D. KL-minimization against deterministic function
- Q7. (Multi-select)** Advantages of DPO over PPO-based RLHF:
- A. Simpler, no rollouts required
 - B. Maximum likelihood training from preferences
 - C. Token-level reward shaping
 - D. End-to-end gradient optimization from preference pairs
- Q8.** What is the key idea behind Group Relative Preference Optimization (GRPO) introduced by DeepSeek?
- A. Using PPO with dynamic temperature scaling
 - B. Token-level reward shaping via causal masking
 - C. Learning from groups of ranked responses rather than only pairs
 - D. Regularizing with KL divergence to the base model
- Q9.** In DPO, which objective is correct?
- A. $\log \pi_{\theta}(y^+) - \log \pi_{\theta}(y^-)$
 - B. $\text{KL}(\pi_{\theta} \parallel \pi_{\text{ref}})$
 - C. $-\log \left(\frac{e^{\beta \log \pi_{\theta}(y^+)}}{e^{\beta \log \pi_{\theta}(y^+)} + e^{\beta \log \pi_{\theta}(y^-)}} \right)$
 - D. Both A and C
- Q10.** Which of the following is a drawback of Direct Preference Optimization (DPO) compared to PPO?
- A. Requires separate reward model training
 - B. Involves costly sampling rollouts
 - C. Lacks token-level reward shaping flexibility
 - D. Suffers from off-policy instability