Backpropagation. Is backpropagation considered biologically plausible? If so, why, if not, why not?

- 1. No. In biological brains, backpropagation would lead to the vanishing gradient problem, in which gradients would go to zero before reaching the necessary depth in the network.
- 2. Yes. Like many Al principles, it is expressly inspired by known biological feedback connections.
- 3. No. Backpropagation requires identical feedback connections for all feedforward connections.
- 4. Yes. This is the role played by dopamine in the mammalian brain.

**Grid Coding.** We can use fMRI to find grid-like coding in the human brain, because:

- 1. When running along a hexagonal grid in a virtual reality environment, grid cells will respond periodically (at 4 Hz) whenever you cross the humps of a grid and this fast periodicity is what fMRI picks up.
- 2. We can't. As explained in class, grid cells are remapped from environment to environment, so this cannot be done in a reasonable fMRI experiment.
- 3. Depending on your direction (modulo 60 degrees) in a space, you will either traverse the activations or silences of all grid cells. So, we analyze fMRI responses contingent on movement direction.
- 4. We can use sophisticated machine learning techniques to find the signatures of grid-like coding, especially in entorhinal cortex.

**Divisive Normalization population receptive field modeling.** There are multiple parameters in the DN pRF model that represent constants in the numerator and denominator in its division:

- 1. Activation constant (B), in the numerator, determines the amount of response compression.
- 2. Normalization constant (D), in the denominator, determines the amount of response compression.
- 3. Normalization constant (D), in the numerator, determines the amount of surround inhibition.
- 4. Activation constant (B), in the numerator, determines the amount of surround inhibition.
- 5. Normalization constant (B), in the denominator, determines the amount of surround inhibition.

Divisive Normalization as a general principle. What 2 statements about divisive normalization are true?

- 1. DN allows responses to be scaled to some background, for example it can rescale local neural responses to light based on the mean light level across the retina
- 2. DN can create winner-take-all suppression between neurons that represent different visual orientations, for example.
- 3. DN can implement the temporal difference learning algorithm for reinforcement learning, but fails to capture the dynamics of the Rescorla-Wagner model
- 4. DN is one of the basic ingredients of recurrent neural network architectures, without which backpropagation-through-time cannot be performed.

**Single-layer perceptron.** There is a specific operation that a single-layer perceptron cannot perform. Which?

- 1. The AND operation
- 2. The NAND operation
- 3. The OR operation
- 4. The XOR operation
- 5. The single-layer perceptron is a universal function approximator: it can perform any operation

Grid Coding. How do responses in Entorhinal cortex (EC) lead to place cell responses in Hippocampus?

- 1. Grid cells form a Fourier basis for the representation of space: place cells could be created through a linear combination of grid cell responses.
- 2. EC already maintains a place cell code, and hippocampus inherits these place cell responses directly.
- 3. EC combines retinotopic activations throughout the brain to create a world-centric representation of space such as those in place cells.
- 4. EC maintains a history of where an animal has been in the past, and by integrating over all past locations EC can create knowledge of where an animal is now.

# Replay. A population of place cells:

- 1. Can replay a sequence of events (a route) more slowly than it happened in actual life so that memory consolidation processes can lay down memories better.
- 2. Can lay down memories, but its activations during navigational decision-making are divorced from choices that an animal makes.

- 3. Can replay a sequence of events (a route) faster than it happened in actual life so that memory consolidation processes connect distant cortical representations.
- 4. Has place cells that are fixed to the environment, and they always have the same place field when we move from environment to environment.

Multi-layer perceptron. There is a specific operation that a multi-layer perceptron cannot perform. Which?

- 1. The AND operation
- 2. The NAND operation
- 3. The OR operation
- 4. The XOR operation
- 5. The multi-layer perceptron is a universal function approximator: it can perform any operation

# Grid Coding. Select the wrong statement. Entorhinal cortex grid cells

- 1. In a single individual differ in their grid spacing (frequency)
- 2. In a single individual differ in their grid offset (phase)
- 3. In a single individual differ in their grid orientation
- 4. Are anchored to world-centered space

#### Tolman.

- 1. Was a behaviorist that, in 1924, discovered extinction in reinforcement learning
- 2. Was a behaviorist that, in 1932, discovered blocking in reinforcement learning, together with Roger Kamin
- 3. Proved, in 1948, that rats maintain a cognitive map of their surroundings even without explicit reward
- 4. Discovered grid cells together with May-Britt Moser, and received the nobel prize in 2014.

# **Encoding models.** The space in which encoding models cannot be formulated:

- 1. Retinotopic visual space
- 2. Tonotopic auditory frequency
- 3. Semantic space
- 4. World-centric space
- 5. There is no space in which encoding models cannot be formulated

# Biological Realism. Find the wrong answer. The basic perceptron model fails to mimic biological neurons in:

- 1. The spatial structure of the dendritic tree, that supports intricate integration of incoming information
- 2. The temporal delay structure of signals coming in on the dendritic tree, again supporting intricate integration of incoming information
- 3. The timing specificity of action potentials
- 4. its fundamental input-output architecture

# Grid fit procedure. Find the wrong answer. Caveats of grid fitting:

- 1. It has limited precision (parameters are fixed on the grid)
- 2. It has predetermined range (meaning you can misspecify this and fail to find the global minimum)
- 3. For high-dimensional parameter spaces, the amount of possible grid elements explodes.
- 4. Cannot be used for least-squares estimation, and has to be used for maximum likelihood estimation

# **Gradient Descent.** Find the wrong answer.

- 1. Gradient descent uses the same rule (delta rule) as that which drives reinforcement learning in for example the Rescorla-Wagner model.
- 2. Gradient descent determines the direction to step in based on the gradient  $d/d\theta J(\theta)$  of our cost function at the present step
- 3. Gradient Descent cannot be used for least-squares estimation, and has to be used for maximum likelihood estimation
- 4. The step size in gradient descent is directly related to the learning rate in a reinforcement learning context.

# Reinforcement learning. The learning rate in a Rescorla-Wagner model:

- 1. Determines the ceiling of the learning: the height of its asymptote
- 2. Depends on the size of the prediction error

- 3. Determines the speed at which learning tends towards its asymptote
- 4. Is larger for a larger reward.

Temporal Difference Learning TD solves this problem of the Rescorla-Wagner model:

- 1. RW cannot function in volatile environments due to its fixed learning rate
- 2. RW does not capture the explore-exploit trade-off
- 3. RW only predicts immediate rewards: no higher-order conditioning
- 4. RW doesn't incorporate actions, TD does

**Signal detection theory.** Imagine a simple signal-in-noise SDT experiment. In the first half, I reward hits but not correct rejections, and in the second half of the experiment, I reward correct rejections but not hits. What will the impact on my participants' behavior be?

- 1. They will be more sensitive in the first half of the experiment, and less sensitive in the second: this will impact D-prime.
- 2. They will adjust their criterion, and be more conservative in the second half while they were more liberal in the first..
- 3. Their percentage correct will be higher for 'noise' (no signal) trials in the second half of the experiment
- 4. Their percentage correct will be higher for 'signal' trials in the second half of the experiment