

## Q1 Gone fishing

5 Points

### Q1.1

1 Point

You go fishing a very large number of times.

The number of fish you catch on a giving fishing trip is a random variable of some distribution.

Which distribution best approximates the total number of fish you catch (throughout all the fishing trips)?

- Binomial
- Geometric
- Poisson
- Exponential
- Normal

#### Explanation

The sum of a set of independent and identically distributed random variables approaches a normal distribution by CLT.

**Q1.2**

**1 Point**

You catch an expected number of 1.5 fish per hour. You can catch a fish at any instant of time.

Which distribution best characterizes the number of fish you catch in one hour of fishing?

- Binomial
- Geometric
- Poisson
- Exponential
- Normal

**Explanation**

You're measuring the discrete number of fish you catch based on the rate at which fish-catching happens in a continuous time interval.

**Q1.3**

**1 Point**

You catch an expected number of 1.5 fish per hour. You can catch a fish at any instant of time.

Which distribution best characterizes how the amount of time it takes to catch a fish after casting the line?

- Binomial
- Geometric
- Poisson
- Exponential
- Normal

**Explanation**

You're measuring the continuous duration of time until a fish is caught.

Q1.4

1 Point

Each day you fish for 2 hours. Assume that the numbers of fish you catch each day are independent.

Which distribution best characterizes the number of days until you catch more than 10 fish in one day?

- Binomial
- Geometric
- Poisson
- Exponential
- Normal

**Explanation**

The probabilities of catching more than 10 fish on each day are independently and identically distributed, and you're measuring the number of trials until success.

Q1.5

1 Point

Each day you fish for 2 hours. Assume that the numbers of fish you catch each day are independent.

Which distribution best characterizes the number of days you catch no fish over a week of fishing?

- Binomial
- Geometric
- Poisson
- Exponential
- Normal

**Explanation**

The probabilities of catching no fish on each day are independently and identically distributed, and you're measuring the total number of successful trials.

## Q2 Normal distribution

3 Points

Let  $X$  be a Normal random variable with mean  $\mu$  and variance  $\sigma^2$ . Which of the following are true statements?

Q2.1

1 Point

$$\frac{X-\mu}{\sigma} \sim N(0, 1)$$

True

False

Explanation

$$\text{Let } Y = \frac{X-\mu}{\sigma}. \mathbb{P}[a \leq Y \leq b] = \mathbb{P}[a\sigma + \mu \leq X \leq b\sigma + \mu] = \frac{1}{\sqrt{2\pi\sigma^2}} \int_{a\sigma+\mu}^{b\sigma+\mu} e^{-(x-\mu)^2/(2\sigma^2)} = \frac{1}{\sqrt{2\pi}} \int_a^b e^{-y^2/2} dy.$$

Q2.2

1 Point

The sum of any two normal random variables is also normally distributed

True

False

Explanation

This is true if and only if the variables are independent. Proof under Corollary 21.1 in Note 21

**Q2.3****1 Point**

The sum of  $n$  i.i.d. random variables follows a normal distribution as  $n$  approaches infinity.

- True  
 False

**Explanation**

Let  $S_n$  be the sum of  $n$  i.i.d random variables each with mean  $\mu$ , variance  $\sigma^2$ . From CLT,  $S_n \sim N(n\mu, n\sigma^2)$ , which is normal.

**Q3 CLT****1 Point**

Suppose  $S_n = X_1 + \dots + X_n$ , where  $X_i$ 's are i.i.d random variables with  $\mathbb{E}[X_i] = \mu$ ,  $\text{Var}(X_i) = \sigma^2$ . As  $n \rightarrow \infty$ , which of the following random variables converges to the standard normal distribution?

- $\frac{S_n}{n}$   
  $\frac{S_n}{n} - \mu$   
  $\frac{S_n - n\mu}{\sigma\sqrt{n}}$   
  $\frac{S_n - n\mu}{\sigma n}$

**Explanation**

Note that by linearity of expectation,  $\mathbb{E}\left[\frac{S_n - n\mu}{\sigma\sqrt{n}}\right] = \frac{1}{\sigma\sqrt{n}}(\mathbb{E}[S_n] - n\mu) = \frac{1}{\sigma\sqrt{n}}(\mathbb{E}[X_1] + \dots + \mathbb{E}[X_n] - n\mu) = 0$ , and since  $X_i$ 's are independent,  $\text{Var}\left(\frac{S_n - n\mu}{\sigma\sqrt{n}}\right) = \frac{1}{\sigma^2 n} \text{Var}(S_n - n\mu) = \frac{1}{\sigma^2 n} \text{Var}(S_n) = \frac{1}{\sigma^2 n} (\text{Var}(X_1) + \dots + \text{Var}(X_n)) = \frac{1}{\sigma^2 n} \cdot \sigma^2 n = 1$ . By CLT,  $\frac{S_n - n\mu}{\sigma\sqrt{n}}$  converges to  $N(0, 1)$  as  $n \rightarrow \infty$ .