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Quiz 1

Q.1 Which of the following statements are true about Bayesian inference?

Max. score: 2; Neg. score: 0; Your score: 0

- For computing the posterior distribution, the denominator in the Bayes rule (the marginal likelihood) is not necessary to compute.
- Fully Bayesian inference with a uniform prior will be equivalent to doing point estimation.
- Uncertainty of the posterior predictive decreases as the amount of training data increases
- Uncertainty of the posterior distribution decreases as the amount of training data increases.

Q.2 Given i.i.d. data $X=\{x_1,x_2,\ldots,x_N\}$ from a distribution $p(x|\theta)$, the following must hold: $\int p(X|\theta)d\theta=1$

Max. score: 1; Neg. score: 0; Your score: 1





false



true

Q.3 Which of the following is true about the concentration parameter vector α of a Dirichlet distribution?

Max. score: 2; Neg. score: 0; Your score: 0

- When the elements of α are very small (close to 0), the Dirichlet becomes similar to a Gaussian distribution
- When the elements of α are very small (close to 0), the Dirichlet becomes similar to a uniform distribution
- When the elements of α are all equal, the Dirichlet becomes peaked (concentrated) towards the center
- When the elements of α are all equal to 1, the Dirichlet becomes similar to a uniform distribution



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	Cross-validation	if held-out data	is available

MAP-II (maximizing the posterior of the hyperparameters)

✓ ✓ MCMC or variational inference

Q.5 Which of the following statements is true regarding conjugacy?

Max. score: 2; Neg. score: 0; Your score: 2

- If likelihood and prior are conjugate and in exponential family, the posterior has a closed form expression
 - Every distribution has an associate conjugate distribution
- If likelihood and prior are conjugate and in exponential family, the posterior predictive has a closed form expression
- If likelihood and prior are conjugate and in exponential family, the marginal likelihood has a closed form expression

Q.6 Posterior predictive distribution will become very similar to the plug-in predictive if

Max. score: 2; Neg. score: 0; Your score: 0

- Number of observations is very small
- **✓** Number of observations is very large
 - If the likelihood and prior are exponential family distributions
 - If we use a uniform prior distribution

Q.7 Large uncertainty in the posterior distribution implies large uncertainty in the posterior predictive distribution.

Max. score: 1; Neg. score: 0; Your score: 0

false

✓ true

Q.8 Any probability distribution can be represented as an exponential family distribution.

Max. score: 1; Neg. score: 0; Your score: 1



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Q.9 For a model with two parameters θ_1 and θ_2 and data X, the overall posterior of the parameters will be

Max. score: 2; Neg. score: 0; Your score: 2

- Defined by two distributions $p(heta_1|X)$ and $p(heta_2|X)$
- $lack ag{Proportional to } p(heta_1, heta_2,X)$
- $lack p(heta_1, heta_2|X)$
 - Defined by two distributions $p(heta_1|X, heta_2)$ and $p(heta_2|X, heta_1)$

Q.10 Using a MAP estimation approach, which of the following are possible?

Max. score: 2; Neg. score: 0; Your score: 2

- We can obtain predictions that are based on posterior averaging over all parameters of all the models under consideration.
- We can obtain predictions that are based on posterior averaging over the parameters of a fixed model.
- ✓ We can prevent overfitting.
- ✓ Predictions can be made only using the plug-in predictive distribution.

Q.11 The regions with very few training observations will typically have a very small variance for the posterior predictive distribution.

Max. score: 1; Neg. score: 0; Your score: 1





false



true

Q.12 Give a brief, intuitive justification as to why the posterior predictive distribution of a probabilistic linear regression model (or, in fact, of any model) would typically have a larger variance than the plug-in predictive distribution. Write your answer in the provided text box only (no file upload).

Max. score: 3; Neg. score: 0; Your score: 0

Your answer:

plug in predictive will take up only the variance of the prior



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Q.13 Assuming two random variables x_a and x_b (assume both to be scalars), having a Gaussian joint distribution, which of the following statements is true?

Max. score: 2; Neg. score: 0; Your score: 2

- $igwedge absolute {f v}_a$ Conditioning one random variable x_a on a given value of another random variable x_b decreases the variance of x_a
 - Conditioning one random variable x_a on a given value of another random variable x_b increases the variance of x_a
- igwedge The mean of the conditional distribution of x_a , given x_b , depends on x_b
 - The mean of the conditional distribution of x_a , given x_b , is independent of x_b

Q.14 Using a prior distribution, which of the following can be achieved?

Max. score: 2; Neg. score: 0; Your score: 0

- Combining it with a likelihood to compute an exact posterior in all cases.
- Regularizing the MLE approach.
- ✓ Pre-specify which parameter values are not likely for the solution
- Compute (exactly/approximate) the marginal likelihood of a model.

Q.15 Consider rolling a dice 10 times and assume the number of times each of the 6 faces show up to be 2,1,3,0,1,3, respectively. Assume a Dirichlet (3,3,3,3,3,3) prior on the probability vector π

Max. score: 2; Neg. score: 0; Your score: 0

- The probability of the next dice roll showing the 4th face, given the full posterior of π is 1/6
- The probability of the next dice roll showing the 4th face, given the MLE solution of π is 0
 - The probability of the next dice roll showing the 4th face is undefined for MLE/MAP as well as for fully Bayesian solution.
 - The probability of the next dice roll showing the 4th face, given the MAP solution of π is 0

Q.16 State two advantages of fully Bayesian inference over MAP estimation. Write your answer in the provided text box only (no file upload).



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gives the variance in parameter estimation Model selection and hyper parameter tuning

Feedback:

Score: 14