

CheatSheet Basics Machine Learning

Chu Duc Thang

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1 Chapter 1

Introduction to the class. Only 1 page, no important information here

2 Chapter 2: Probability

1. Sample space/outcome space vs event space:
 - Sample space: Ω
 - Event space: Subset of sample space, ex: powerset (discrete), Borel Field (continuous)
2. Discrete vs Continuous RV
 - Discrete: $\{\}$, \mathbb{N} , words
 - Continuous: \mathbb{R} , \mathbb{R}^k
3. Probability mass function (pmf) vs probability density function (pdf)
 - Pmf: $\Omega \rightarrow [0, 1]$
 - Pdf: $\Omega \rightarrow [0, \infty)$, no singleton event, can be > 1
4. Special Distribution
 - Discrete: Uniform (n - #outcomes), Poisson (α - histogram/likely), Bernoulli (p - success)
 - Continuous: Gamma (α, β), Uniform (a, b), Normal(μ, σ), Exponential (α)
5. Marginal vs Conditional Distribution
 - Marginal: $p(x) = \sum_{y \in Y} p(x, y)$
 - Conditional: $p(x|y) = \frac{p(y|x)p(x)}{p(y)}$ or $p(x, y, z) = p(x|y, z)p(y|z)p(z)$
6. Expected value vs Conditional Expected value vs Variance

- $E = \sum_{x \in X} xp(x)$
- $E[X|Y] = \sum_{x \in X} xp(x|y)$
- $\text{Var} = E[(X - E[X])^2]$ or $E[X^2] - E[X]^2$
- Properties of E: $E[c] = c$, $E[cX] = cE[X]$, $E[X + Y] = E[X] + E[Y]$, $E[XY] = E[X]E[Y]$ (independence), $E[E[Y|X]] = E[Y]$
- Properties of Var: $\text{Var}[c] = 0$, $\text{Var}[cX] = c^2\text{Var}[X]$, $\text{Var}[X + Y] = \text{Var}[X] + \text{Var}[Y] + 2\text{Cov}(X, Y)$

7. Covariance vs Correlation

- $\text{Cov} = E[XY] - E[X]E[Y]$
- $\text{Corr} = \frac{\text{Cov}(x, y)}{\sqrt{\text{Var}(x)}\sqrt{\text{Var}(y)}}$
- Note: $-1 \leq \text{Corr} \leq 1$, but Cov is unbounded

8. Independence vs Conditional Independence

- $P(X, Y) = P(X)P(Y)$
- $P(X, Y|Z) = P(X|Z)P(Y|Z)$

3 Chapter 3: Estimator

4 Chapter 4: Optimization

5 Chapter 5: MAP/MLE/Bayesian

6 Chapter 6: Optimal predictor

7 Chapter 7: Linear/Polynomial Regression

8 Chapter 8: Generalization Error

9 Chapter 9: Regularization

10 Chapter 10: Classification