

Probability Theory - II

EE698V - Machine Learning for Signal Processing

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Announcements

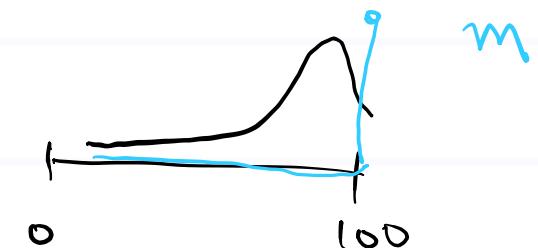
- Quiz on Saturday 8-9PM in L-7
- Should we have a discussion hour tomorrow?
 - 12-1PM in ACES. Others can come 1-2PM in my office.
- If you are facing any problems with any aspect of the course, please do let me know

References

- PRML Section: 1.2 (**highly recommended**)

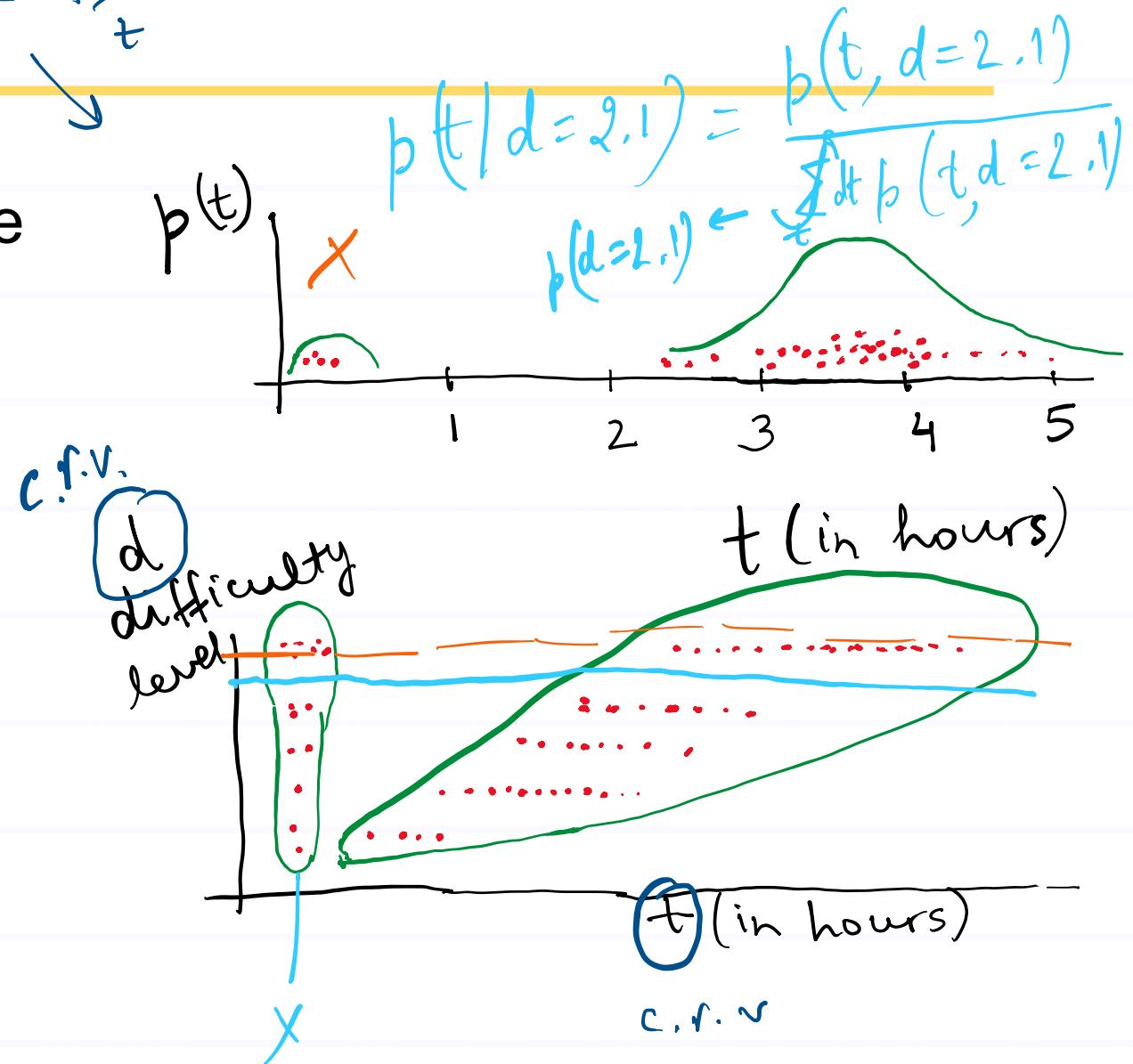
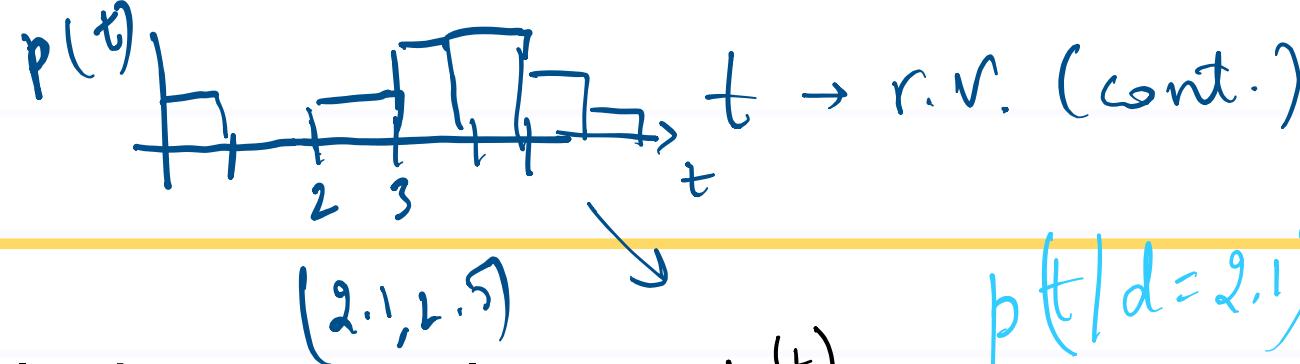
Intuitive Understanding

- Bayesian view
 - My belief on the value of a random variable
 - E.g.,
 - How many marks will I get in the exam?
 - How long will I take to complete my assignment? → t
- You get lumps of samples; approximate with a function



Examples

- How long shall I take to complete this assignment?
- How long do people take to complete assignments ...
- $p(t, d)$
- Limited observation capacity, so we use intuition to build models

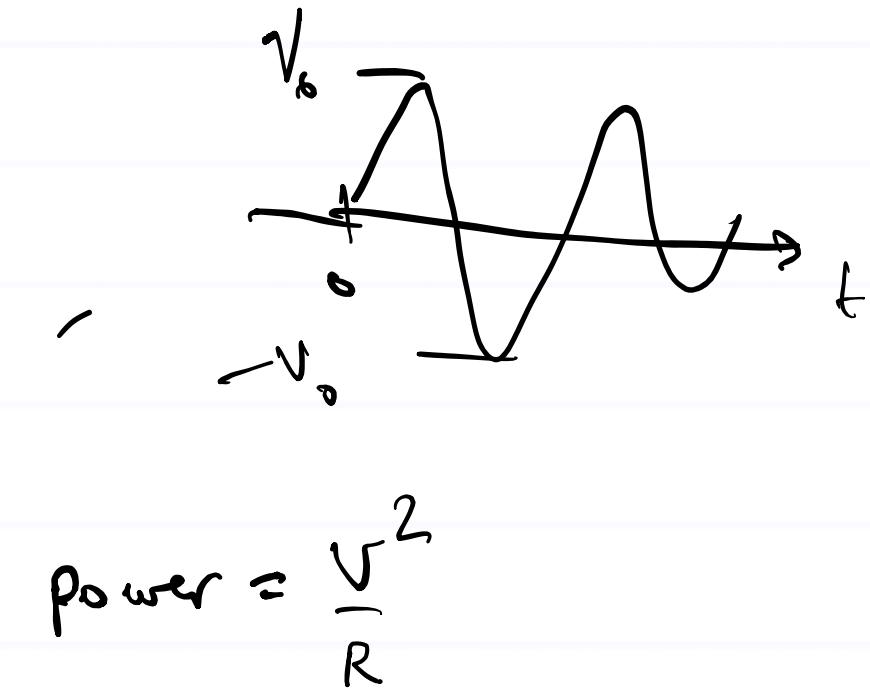
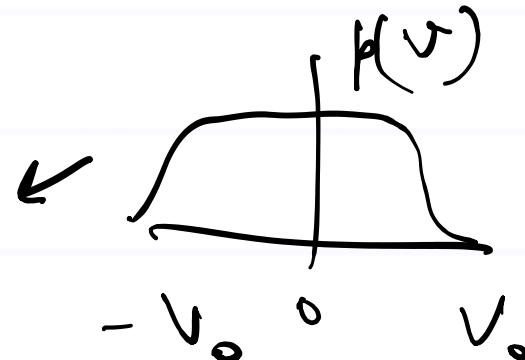
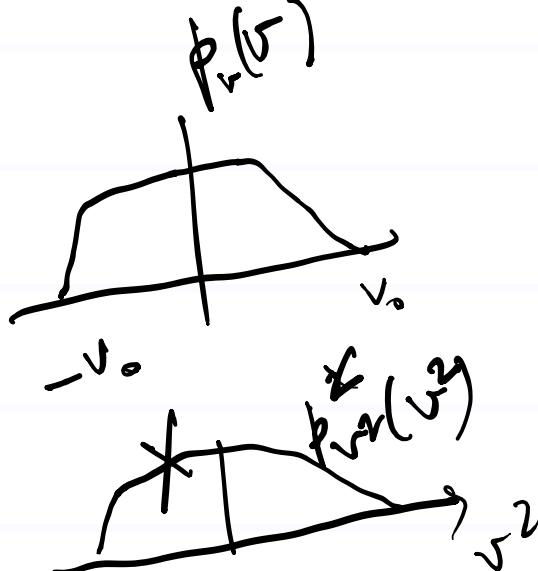


Marginalization, Conditional prob., etc.

- $p(x) = \int p(x, y) dy$ Sum Rule
- $p(x, y) = p(x|y)p(y)$ Product Rule
- $p(y|x) = \frac{p(x|y)p(y)}{p(x)}$ Bayes Rule

Transformation of Variables

- Given $p(x)$, what is $p(x^2)$?
- For voltage readings across a resistor, $p(\text{voltage})$ is known, what is $p(\text{power})$?



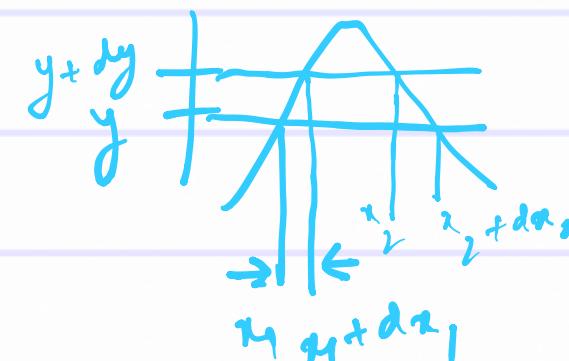
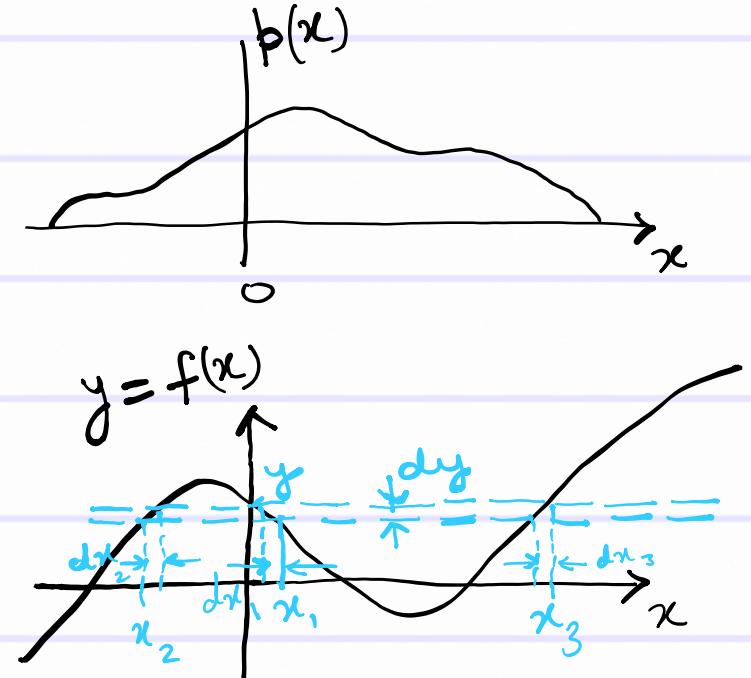
Transformation of Variables

- Given $p_x(x)$ and $y = f(x)$, find $p_y(y)$
- At some x_1 , $y = f(x_1)$
- But there may be other x too that map to same y ,
say $f(x_i) = y$ for $i = 1, 2, \dots, N$

$$p_y(y) dy = p_x(x_1) dx_1 + p_x(x_2) dx_2 + \dots$$

$y \in (y, y+dy)$
 $x \in (x_1, x_1+dx_1)$

$$p_y(y) = \sum_i \frac{p_x(x)}{\left| \frac{dy}{dx} \right|} \Big|_{x_i}$$



$$dx_2 \neq dx_1$$

Expectations

- What is the expected (average) width of a rose? How much do they deviate from this average value?
- What is the expected area of the top of a rose?
- What is the expected (average) width of all roses with height greater than 2cm?

$$A = \pi r^2$$

$$\frac{1}{A} (\pi r^2)$$

w, h

$$E [x(w | h > 2)]$$

Expectations

- Expected value of $f(x)$ Mean=μ
- $\mathbb{E}[f(x)] = \int f(x)p(x)dx$
- Variance of $f(x)$ Variance=σ²
- $\text{var}[f(x)] = \mathbb{E}[(f(x) - \mathbb{E}[f(x)])^2]$ Standard Deviation=σ

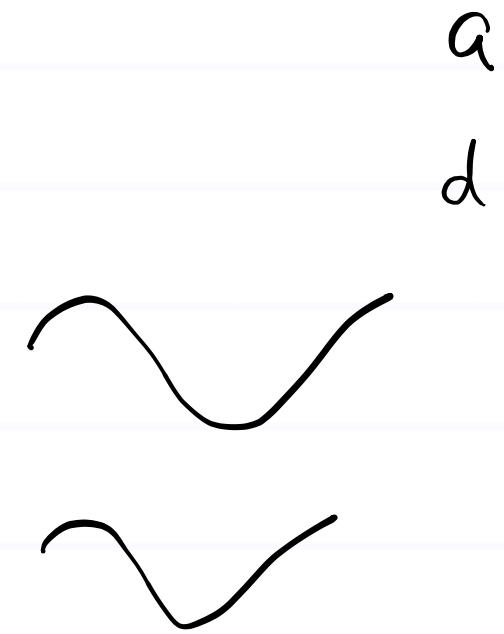
Conditional Expectation

- Conditional Expectation

- $\mathbb{E}_x[f(x)|y] = \int f(x)p(x|y)dx$

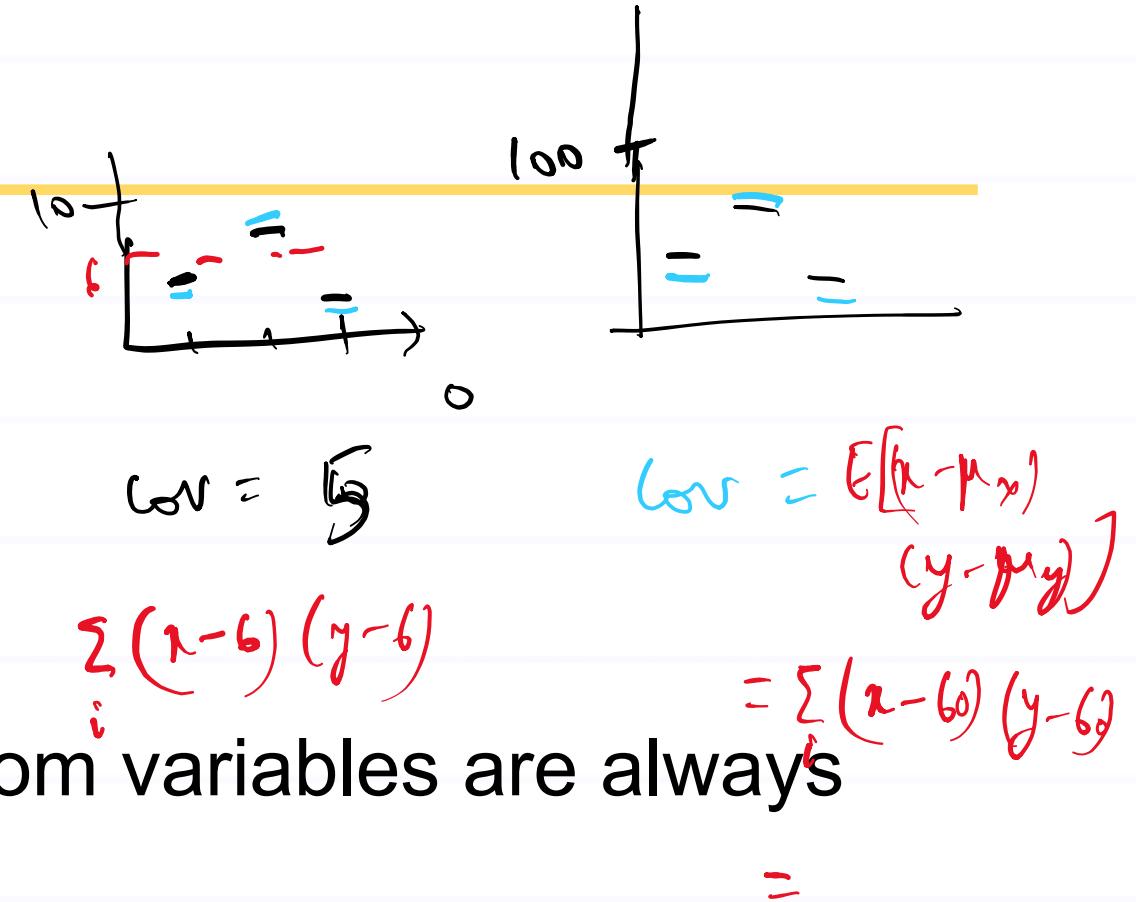
- Covariance

- $\text{cov}[x, y] = \mathbb{E}[(x - \mathbb{E}[x])(y^T - \mathbb{E}[y^T])]$



Correlation

- $\rho_{x,y} = \text{corr}(x, y) = \frac{\text{cov}[x, y]}{\sigma_x \sigma_y}$ ✓
- Uncorrelated $\Rightarrow \rho_{x,y} = 0$
- Prove that two independent random variables are always uncorrelated
- Are two uncorrelated random variables always independent? Prove your answer



$$p(x, y) = p(x) p(y)$$

$$\text{corr} = \frac{1}{\sigma_x \sigma_y} E[(x - \mu_x)(y - \mu_y)]$$

$$\begin{aligned} \text{num.} &= \iint (x - \mu_x)(y - \mu_y) \underbrace{p(x, y)}_{p(x) p(y)} dx dy \\ &= \int (x - \mu_x) p(x) dx \quad \underbrace{\int (y - \mu_y) p(y) dy}_{\mu_x - \mu_y} \end{aligned}$$

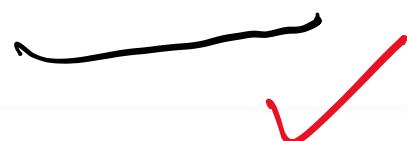
Samples vs pdf

- Samples $\{s_1, s_2, s_3, \dots, s_{50}\}$

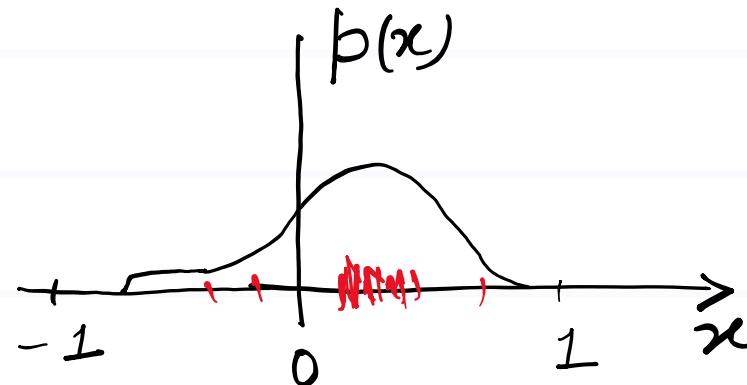
obtained by measuring x

- $E[x] = \int x p(x) dx = \sum_i x_i p(x_i)$

- $E[x] \approx \frac{1}{50} \sum_i s_i$



No $p(x)$ here



$$E[x] = \sum_i s_i p(s_i)$$

S are
samples
from
discr. r.v. X

Samples vs pdf

- $E[f(x)] = \int f(x)p(x)dx$
- $E[f(x)] \approx \frac{1}{50} \sum_i f(s_i)$

Standard distributions

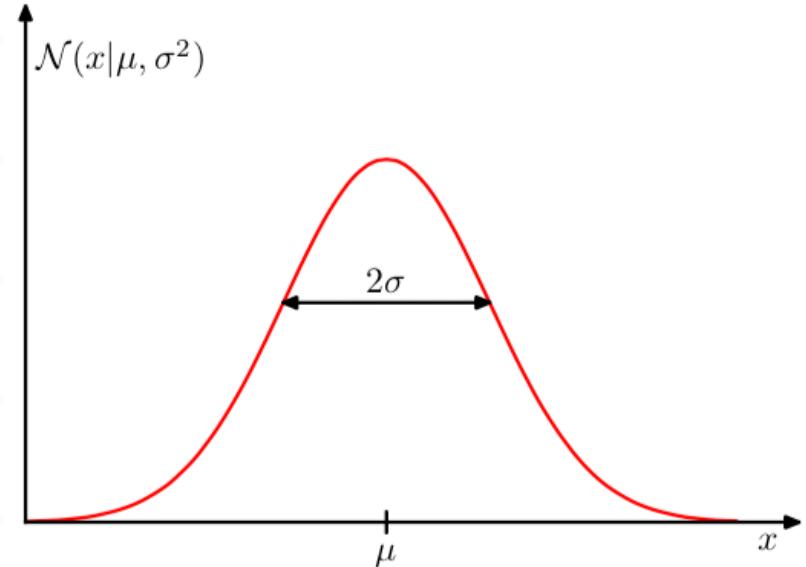
- Known or easy-to-compute properties. E.g.,
 - integration
 - marginalization
 - derivatives
- Examples
 - Bernoulli, Multinomial, etc.
 - Gaussian, Laplacian, etc.

Gaussian Distribution

$$\mathcal{N}(x|\mu, \sigma^2) = \frac{1}{(2\pi\sigma^2)^{1/2}} \exp\left\{-\frac{1}{2\sigma^2}(x - \mu)^2\right\}$$

$$\mathbb{E}[x] = \int_{-\infty}^{\infty} \mathcal{N}(x|\mu, \sigma^2) x \, dx = \mu.$$

$$\mathbb{E}[x^2] = \int_{-\infty}^{\infty} \mathcal{N}(x|\mu, \sigma^2) x^2 \, dx = \mu^2 + \sigma^2.$$



Estimating the parameters

- I have samples. I want to model it with Gaussian distribution. How do I estimate the model parameters?
- Random variable is t
- I want to maximize the probability of observed samples $\{s_1, s_2, \dots, s_N\}$

