# A primer in statistics – Conditional, Joint and marginal distributions

Sensor fusion & nonlinear filtering

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#### CONDITIONAL DISTRIBUTIONS

 Conditional distributions are indispensable components in sensor fusion, filtering and Bayesian estimation in general.

## Conditional distribution (product rule)

- Let x and z be two random variables with the joint pdf p(x, z).
- The conditional density function, p(z|x), is defined through

The conditional density function, 
$$p(z|x)$$
, is defined through 
$$p(x,z) = p(z|x)p(x),$$
 and if  $p(x) \neq 0$  this implies that 
$$p(z|x) = \frac{p(x,z)}{p(x)}.$$

• Interpretation: p(z|x) describes the distribution of z given that x is known.

### **CONDITIONAL DISTRIBUTIONS**

## **Example: Candy problem**

- Every day Sara decides how many pieces of candy she can have for an after lunch snack.
- With 40% probability she tosses a coin, heads means 1 piece and tails means 0 pieces
- With 60% probability she throws a dice (number on the dice = number of candies).

• If z denotes number of candies she eats
$$\Pr\{z=i | \text{Sara tosses a coin}\} = \begin{cases} 0.5 & \text{if } i=0,1\\ 0 & \text{otherwise} \end{cases}$$

$$\Pr\{z=i | \text{Sara throws a dice}\} = \begin{cases} 1/6 & \text{if } i=1,2,...,6\\ 0 & \text{otherwise} \end{cases}$$

#### LAW OF TOTAL PROBABILITY

 Many important results in non-linear filtering is obtained from the law of total probability.

## Law of total probability (sum rule)

• If x takes values in a set  $S_x$ , the law of total probability states that

Discrete: 
$$\Pr\{z\} = \sum_{x \in S_x} \Pr\{x, z\} = \sum_{x \in S_x} \Pr\{z | x\} \Pr\{x\}$$

Continuous: 
$$p(z) = \int_{x \in S_x} p(x, z) dx = \int_{x \in S_x} p(z|x)p(x) dx$$

## LAW OF TOTAL PROBABILITY

# **Example: Candy pmf**

• To calculate the pmf for the number of candies we use

$$\Pr\{z\} = \sum_{x \in S_x} \Pr\{z \mid x\} \Pr\{x\},$$

where x is either 'Sara tosses a coin' or 'Sara throws a dice'.

