

# Today

Random Variables.

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  - ▶ Inclusion/Exclusion:  $Pr[A \cup B] = Pr[A] + Pr[B] - Pr[A \cap B]$ .
  - ▶ Simple Total Probability:  $Pr[B] = Pr[A \cap B] + Pr[\bar{A} \cap B]$ .
  - ▶ Complement:  $Pr[\bar{A}] = 1 - Pr[A]$ .
  - ▶ Union Bound. Total Probability.
- ▶ **Conditional Probability:**  $Pr[A|B] = \frac{Pr[A \cap B]}{Pr[B]}$
- ▶ **Bayes' Rule:**  $Pr[A_m|B] = p_m q_m / (p_1 q_1 + \dots + p_M q_M)$ .

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1. Random Variables.
2. Expectation
3. Distributions.

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The number is a (known) function of the outcome.

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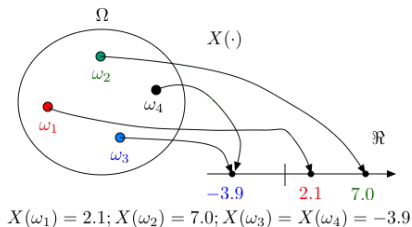
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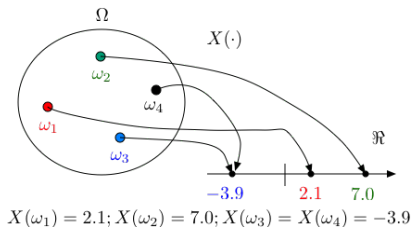
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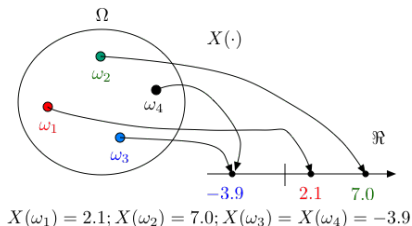


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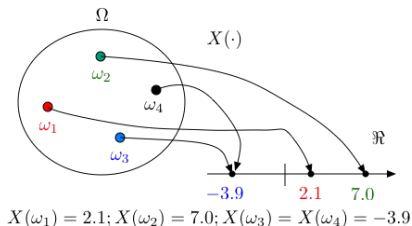


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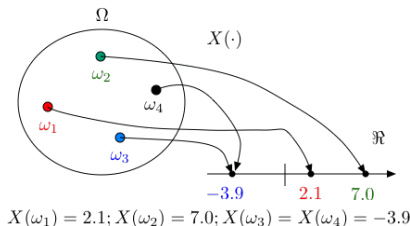
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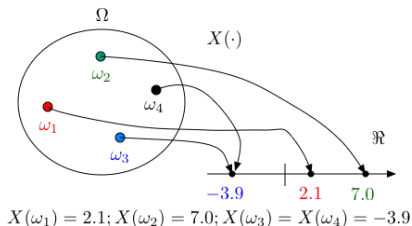
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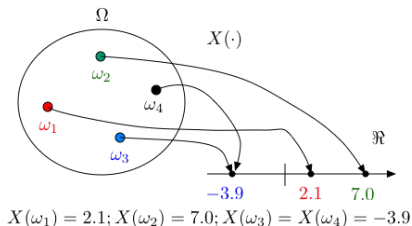
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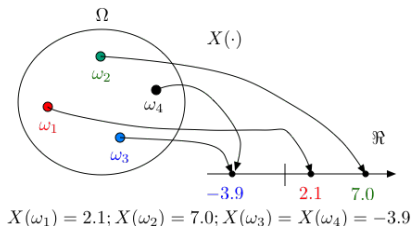
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Note: **Random variable induces partition:**

$$A_y = \{\omega \in \Omega : X(\omega) = y\} = X^{-1}(y)$$

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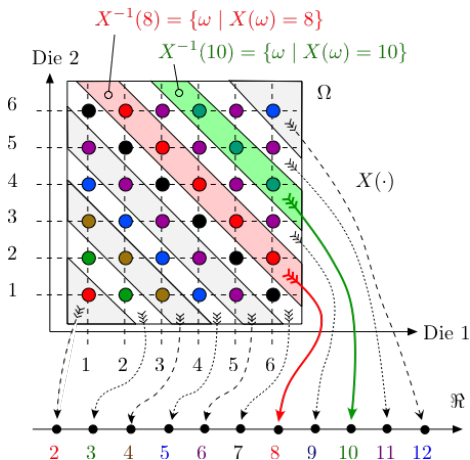
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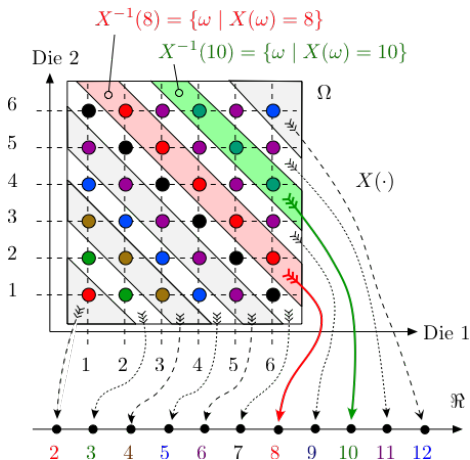
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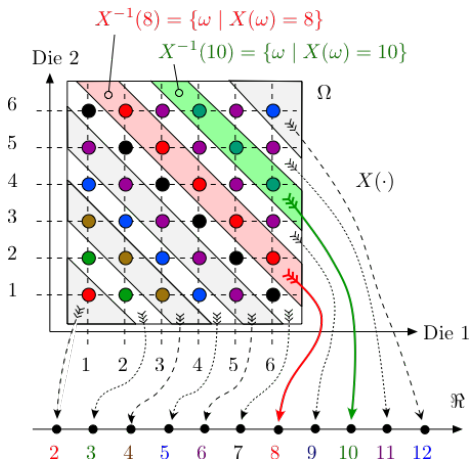
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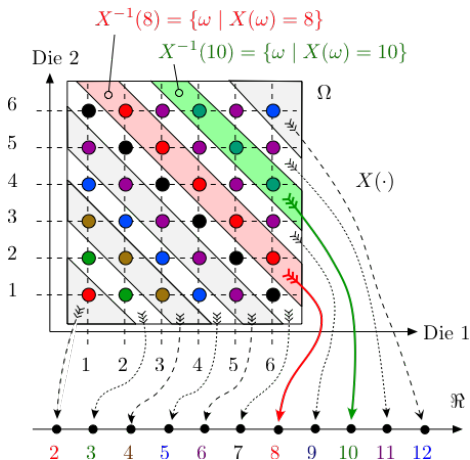
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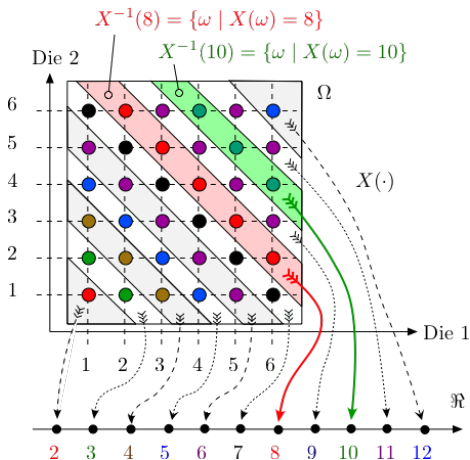
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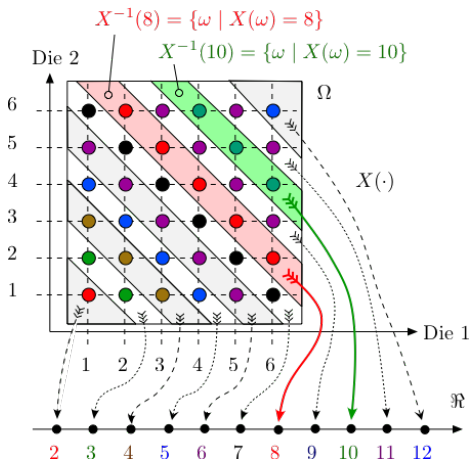
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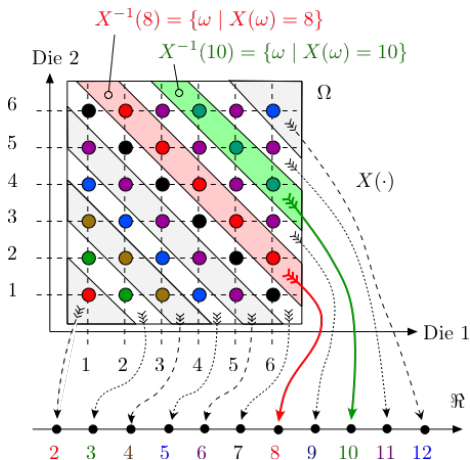
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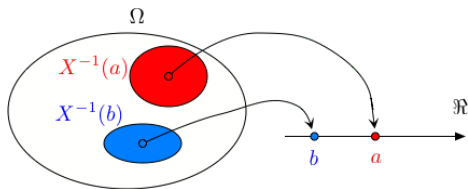
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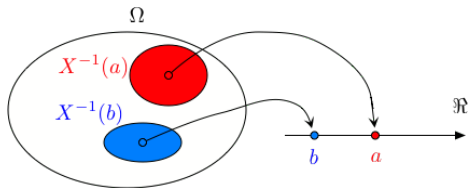
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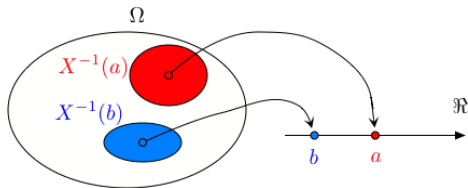


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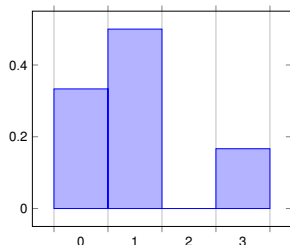
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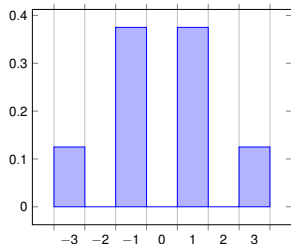
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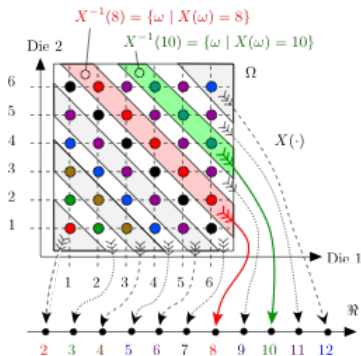


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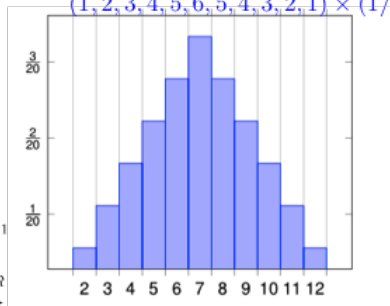
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$(1, 2, 3, 4, 5, 6, 5, 4, 3, 2, 1) \times (1/36)$





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The subjectivist(bayesian) interpretation of  $E[X]$  is less obvious.

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Distributive property of multiplication over addition.

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“Average score” of the  $n$  students: add scores and divide by  $n$ :

$$\text{Average} = \frac{X(1) + X(1) + \dots + X(n)}{n}.$$

Experiment: choose a student uniformly at random.

Uniform sample space:  $\Omega = \{1, 2, \dots, n\}$ ,  $Pr[\omega] = 1/n$ , for all  $\omega$ .

Random Variable: midterm score:  $X(\omega)$ .

Expectation:

$$E(X) = \sum_{\omega} X(\omega) Pr[\omega] = \sum_{\omega} X(\omega) \frac{1}{n}.$$

Hence,

$$\text{Average} = E(X).$$

This holds for a **uniform** probability space.

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Let's cover some.

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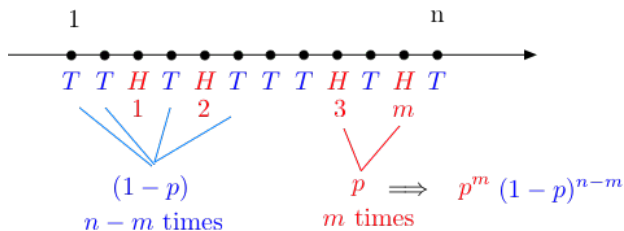
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Probability of “ $X = i$ ” is sum of  $Pr[\omega]$ ,  $\omega \in “X = i”$ .

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# The binomial distribution.



$\binom{n}{m}$  outcomes with  $m$  Hs and  $n-m$  Ts

$$\Rightarrow Pr[X = m] = \binom{n}{m} p^m (1-p)^{n-m}$$

## Error channel and...

A packet is corrupted with probability  $p$ .



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Also distribution in polling, experiments, etc.

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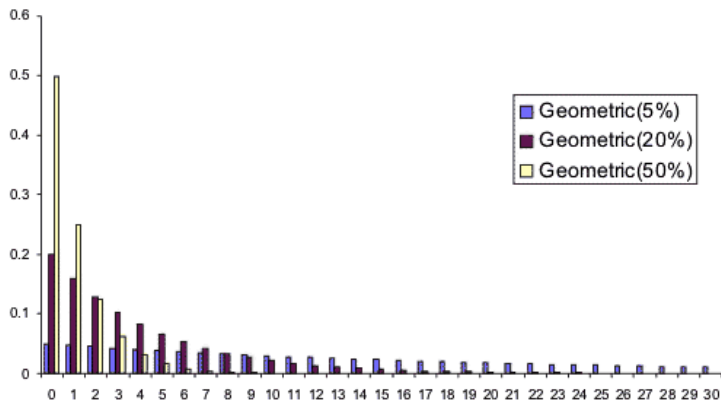
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