

Lecture 3.1, MATH-57091 Probability and Statistics for High-School Teachers.

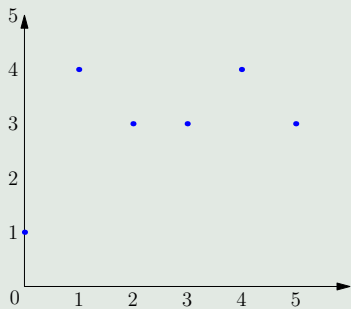
Artem Zvavitch

Department of Mathematical Sciences, Kent State University

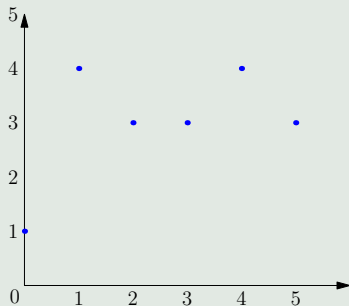
September, 8 - 12, 2014.

A bit different look at functions.

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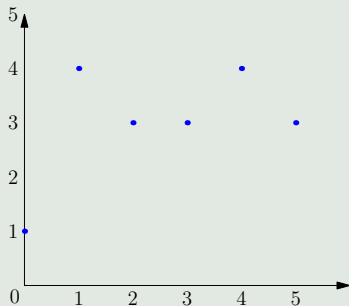
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$f(x)$	How often
1	1
3	3
4	2

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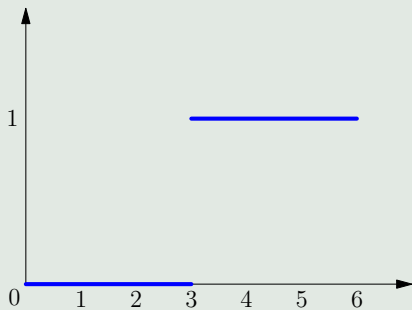


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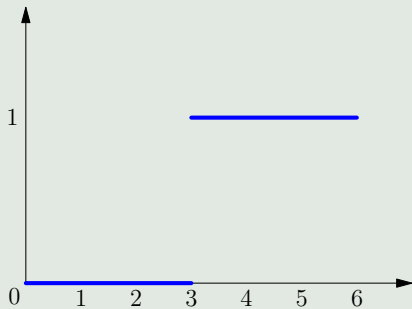
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f	Frequency/Probability
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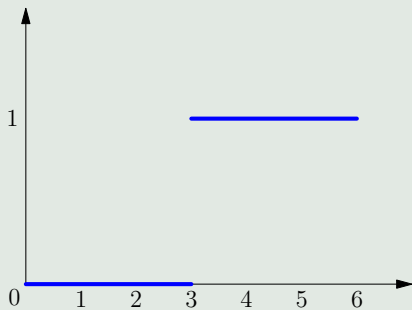


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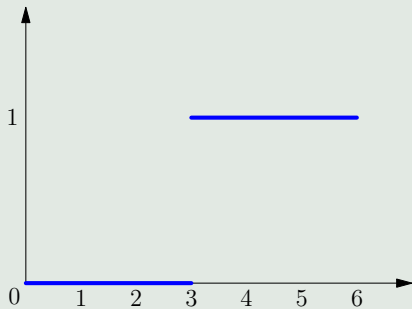
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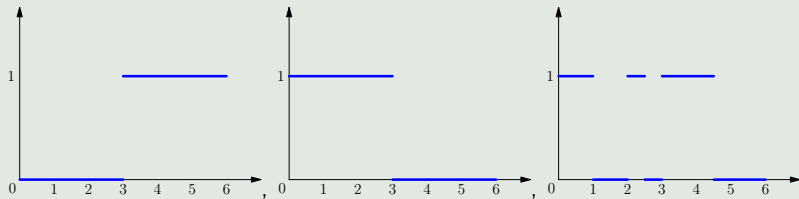
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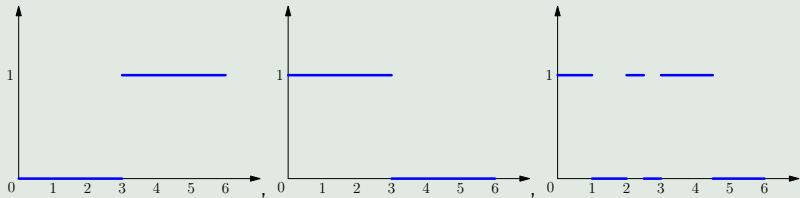
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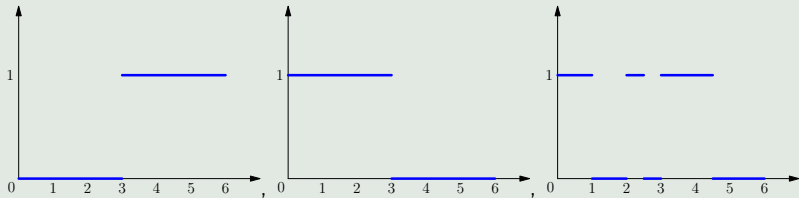


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Random variable

When a probability experiment is performed, often we are not interested in all the details of the experimental result, but rather are interested in the value of some numerical quantity. This gives the definition of **random variable**: a real valued function on the sample space. Since the value of random variable is determined by the outcome of the experiment, we may assign probabilities of its possible values.

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$$P(X = 2) = P((1, 1)) = \frac{1}{36},$$

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We also may check that

$$\sum_{i=1}^{\infty} (1 - p)^{i-1} p = p \sum_{i=1}^{\infty} (1 - p)^{i-1} = p \frac{1}{1 - (1 - p)} = 1$$

Distribution function of random variable X : $F_X(t) = P(X \leq t)$

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It easy (simple logic) to check that $F_X(t)$ is not decreasing function (i.e. $F_X(t_1) \leq F_X(t_2)$ for $t_1 \leq t_2$). More over (check!):

$$P(a < X \leq b) = F_X(b) - F_X(a).$$

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or

$$.4 + .1 + P(X = 3) = 1$$

and $P(X = 3) = .5$

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$$\mathbb{E}Y = \sum_{i=1}^6 y_i P(Y = y_i) = 1 * \frac{1}{6} + 2 * \frac{1}{6} + 3 * \frac{1}{6} + 4 * \frac{1}{6} + 5 * \frac{1}{6} + 6 * \frac{1}{6} = 3.5$$

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Solution: We know (see the beginning of this lecture) that

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$$P(X = 6) = \frac{5}{36},$$

$$P(X = 7) = \frac{6}{36},$$

$$P(X = 8) = \frac{5}{36},$$

$$P(X = 9) = \frac{4}{36},$$

$$P(X = 10) = \frac{3}{36},$$

$$P(X = 11) = \frac{2}{36},$$

$$P(X = 12) = 1/36.$$

Let X denote the random variable that is defined as the sum of two fair dice, find $\mathbb{E}X$.

Solution: We know (see the beginning of this lecture) that

$$P(X = 2) = \frac{1}{36},$$

$$P(X = 3) = \frac{2}{36},$$

$$P(X = 4) = \frac{3}{36},$$

$$P(X = 5) = \frac{4}{36},$$

$$P(X = 6) = \frac{5}{36},$$

$$P(X = 7) = \frac{6}{36},$$

$$P(X = 8) = \frac{5}{36},$$

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the direct computations give us $\mathbb{E}X = \sum_{i=1}^{12} x_i P(X = x_i) = 7$.

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The above computations are direct a correct but there is a much simpler way to do it! Using the properties of expected value. We will do it during next lecture!