

# Chebyshev's Theorem

Theorem: The probability that any random variable  $X$  will assume a value within  $k$  standard deviations of its mean is at least  $1 - 1/k^2$ .  
In other words,

$$P(\mu - k\sigma < X < \mu + k\sigma) \geq 1 - \frac{1}{k^2}$$

Notice: a) This is regardless of the specific distribution  $f(x)$

b) This is a lower bound not an exact equality.

Proof:  $\sigma^2 = \int_{-\infty}^{\infty} (x - \mu)^2 f(x) dx$

$$= \underbrace{\int_{-\infty}^{\mu - k\sigma} (x - \mu)^2 f(x) dx}_{A \geq 0} + \underbrace{\int_{\mu - k\sigma}^{\mu + k\sigma} (x - \mu)^2 f(x) dx}_{B \geq 0} + \underbrace{\int_{\mu + k\sigma}^{\infty} (x - \mu)^2 f(x) dx}_{C \geq 0}$$

$$\geq A + C$$

Now notice that when  $x \leq \mu - k\sigma$  or  $x \geq \mu + k\sigma$  we have  $|x - \mu| \geq k\sigma \Rightarrow (x - \mu)^2 \geq k^2\sigma^2$

so  $\sigma^2 \geq A + C \geq \int_{-\infty}^{\mu - k\sigma} k^2\sigma^2 f(x) dx + \int_{\mu + k\sigma}^{\infty} k^2\sigma^2 f(x) dx$

$$\sigma^2 \geq k^2\sigma^2 \left( \int_{-\infty}^{\mu - k\sigma} f(x) dx + \int_{\mu + k\sigma}^{\infty} f(x) dx \right)$$

$$1 - P(\mu - k\sigma < X < \mu + k\sigma)$$

$$\frac{1}{k^2} \geq 1 - P(\mu - k\sigma < X < \mu + k\sigma)$$

$$P(\mu - k\sigma < X < \mu + k\sigma) \geq 1 - 1/k^2$$

\* Chebyshev's theorem is useful when trying to compute a lower limit for a probability when the exact distribution of  $X$  is not known but the mean and variance are known.


\* This happens in statistics = we make multiple observations of the outcome of a random variable  $X$  so we can estimate the mean and variance, but we don't know the exact  $f(x)$ .

Example =  $X$ ,  $\mu_x = 8$ ,  $\sigma_x^2 = 9$ , unknown  $f(x)$

center a)  $P(-4 < X < 20) = P(\mu_x - 4\sigma_x < X < \mu_x + 4\sigma_x)$   $k=4$

  $\geq 1 - \frac{1}{16}$

tails b)  $P(|X - 8| \geq 6) = 1 - P(\mu_x - 2\sigma_x < X < \mu_x + 2\sigma_x)$

  $\geq 1 - 1/4$   
so  $P(|X - 8| \geq 6) \leq 1/4$

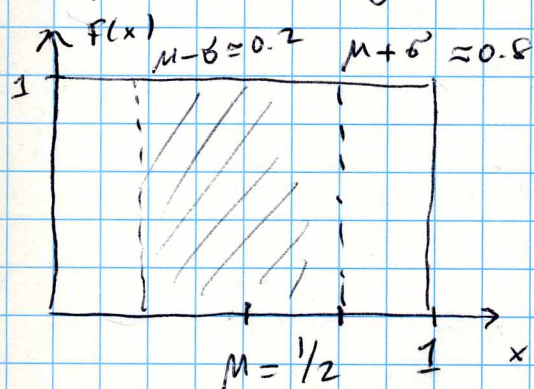
Example = How good is the bound for a normal distribution?

From Table A.3  $P(-3 < Z < 3) = 1 - 2 \times 0.0044$   
 $= 0.9912$

Chebyshev  $P(-3 < Z < 3) \geq 1 - \frac{1}{9} \approx 0.89$

Not too bad. (for  $f(x)$  with long tails like normal dist)

Example = How good is the bound for uniform distribution?



$P(\mu - \sigma < X < \mu + \sigma) = 0.6$

Chebyshev:

$P(\mu - \sigma < X < \mu + \sigma) \geq 1 - 1 = 0$

Not useful at all.

$\sigma^2 = 1/12$  (shown earlier)