## Normal distribution

Some technical remarks

- Standard normal:  $Z \sim \mathcal{N}(0, 1)$ .
- Interesting properties of the log of the pdf of the normal (not to be confused with the 'lognormal'!):

$$\varphi(x) \equiv -\ln f(x) = \frac{(x-\mu)^2}{2\sigma^2} + k$$
  

$$\varphi_{min} = k$$
  

$$\Delta \varphi^{\left(\frac{1}{2}\right)} = \varphi(\mu \pm \sigma) - \varphi_{min} = \frac{1}{2}$$
  

$$\frac{d\varphi}{dx} = \frac{x-\mu}{\sigma^2} \xrightarrow[==0]{} x_m = \mu$$
  

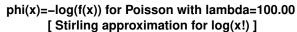
$$\frac{d^2\varphi}{d^2x} = \frac{1}{\sigma^2}$$

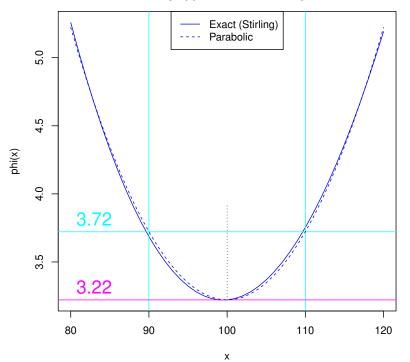
- ⇒ Useful 'tricks' to evaluate mean and variance of distributions "assumed to be almost normal".
- Beware when they become rules! (Or prescriptions, or even 'principles'...)

## (Quite academic) example of the 'Gaussian tricks'

In the case of a Poisson distribution, we can calculate  $\varphi(x)$ , making x continuous and using Stirling's approximation:

$$\varphi(x) = -\ln\left(\frac{e^{-\lambda}\lambda^{x}}{x!}\right)$$
  
=  $+\lambda - x\ln\lambda + \ln(x!)$   
 $\approx \lambda - x\ln\lambda + x\ln x - x + \frac{1}{2}\ln x + \frac{1}{2}\ln 2\pi$ 





A numerical example ( $\lambda = 100$ ), using E[X] =  $\sigma^2 = \lambda$ , and comparing with a parabolic function around the minimum.

## Exercise on the 'Gaussian trick' applied to the Poisson distribution Evaluation of E[X] and $\sigma^2(X)$ , for large $\lambda$ and treating x as continuous

$$\varphi(x) = \lambda - x \ln \lambda + x \ln x - x + \frac{1}{2} \ln x + \frac{1}{2} \ln 2\pi \qquad (1)$$
  
$$\frac{d\varphi}{dx} = \qquad (2)$$
  
$$\frac{d^2 \varphi}{d^2 x} = \qquad (3)$$

 $\lambda = 100$ :

- 1. 'estimate' E[X] mimizing (1);
- 2. 'estimate' E[X] from the root of (2), i.e. from  $d\varphi/dx = 0$ ;
- 3. then 'estimate'  $\sigma^2(X)$  from (3), after having 'got' E[X] from the previous items.
- 4. Finally 'estimate'  $\sigma(X)$  using the ' $\Delta \varphi$  trick'.