Duke University Department of Physics

Physics 391

Fall Term 2010

WUN2K FOR LECTURE 2

These are notes summarizing the main concepts you need to understand and be able to apply.

- For a situation in which one assumes some underlying "truth" description, but for which specific results can vary, the probability of a particular result is the ratio of the number of times that result occurs to the total number of occurrences. A probability is a number in the range 0 to 1, and the distribution of probabilities must be normalized, *i.e.* $\sum_{\text{possible results i}} P(i) = 1$ for discrete results, and $\int_{-\infty}^{\infty} P(x) dx$ for continuous results.
- Rules of probability:
 - The probability of A OR B occurring, P(A + B), must satisfy $P(A + B) \leq P(A) + P(B)$, where the equality holds if A and B are independent.
 - The probability of A AND B occurring, P(AB), is given by P(AB) = P(A|B)P(B) = P(B|A)P(A), where the notation P(A|B) means "the probability of A, given B". This relation is also known as "Bayes' Law".
- The binomial distribution is $B(x; n, p) = \binom{n}{x} p^x q^{n-x}$: this is the probability of x "successes" in n trials, where the probability of success is p, and q = 1 p. (Here $\binom{n}{x} = \frac{n!}{x!(n-x)!}$.) The mean of a binomial is $\mu = np$ and the variance is $\sigma^2 = np(1-p)$.
- The Poisson distribution is the limit of the binomial for the case $\mu \ll n$ and $p \ll 1$: $\mathcal{P}(x;\mu) = \frac{\mu^x}{x!}e^{-\mu}$. Its mean is μ and its variance is also

 $\sigma^2=\mu;$ it's described by the single parameter $\mu.$ The Poisson is often useful for counting statistics.

• The Gaussian distribution is the limit of the binomial for np >> 1, and also the limit of the Poisson for $\mu >> 1$: it's $G(x; \mu, \sigma) = \frac{1}{\sigma\sqrt{2\pi}}e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$, with mean μ and variance σ^2 . Some useful properties:

 $G(\mu\pm\sigma;\mu,\sigma)=G(\mu;\mu,\sigma)/e.$

The "full width half max (FWHM)", for which $G(\mu \pm \frac{\Gamma}{2}; \mu, \sigma) = \frac{G(\mu;\mu,\sigma)}{2}$ is $\Gamma = 2.354\sigma$.

The area between $\mu - \sigma$ and $\mu + \sigma$ is 68.3% of the total area; between $\mu - 2\sigma$ and $\mu + 2\sigma$ is 95.5%; and between $\mu - 3\sigma$ and $\mu + 3\sigma$ is 99.7%. Beware of "non-Gaussian tails" when going out many sigmas, though!