

**Homework 8, Solution sketches**

Durrett, 3.3.2.(i) By Fubini,

$$\begin{aligned} \frac{1}{2T} \int_{-T}^T e^{-ita} \varphi(t) dt &= \frac{1}{2T} \int_{-T}^T dt e^{-ita} \int_{\mathbb{R}} e^{itx} d\mu(x) \\ &= \int_{\mathbb{R}} d\mu(x) \frac{1}{2T} \int_{-T}^T e^{it(x-a)} dt \\ &= \int_{\mathbb{R}} d\mu(x) \frac{1}{2T} \int_{-T}^T \cos(t(x-a)) dt. \end{aligned}$$

Let

$$I(x, T) = \frac{1}{2T} \int_{-T}^T \cos(t(x-a)) dt.$$

Then clearly  $|I(x, T)| \leq 1$  for all  $x$  and  $t$ . Moreover, by computing the integral we see that, as  $T \rightarrow \infty$ ,  $I(x, T) \rightarrow 1$  if  $x = a$  and  $I(x, T) \rightarrow 0$  otherwise. Now use DCT.

Durrett, 3.3.3. As  $\varphi_{X-Y} = \varphi_X \varphi_{-Y} = \varphi_X \overline{\varphi_Y} = \varphi_X \overline{\varphi_X} = |\varphi_X|^2$ , the first equality follows from the previous problem. To prove the second equality, let  $x_k$  be an enumeration of points  $x$  with  $P(X = x) > 0$ . Then

$$\begin{aligned} P(X = Y) &= \int_{\mathbb{R}} d\mu(x) \int_{\mathbb{R}} 1_{\{x=y\}} d\mu(y) \\ &= \int_{\mathbb{R}} P(X = x) d\mu(x) \\ &= \int_{\mathbb{R}} \sum_k P(X = x_k) 1_{\{x_k\}}(x) d\mu(x) \\ &= \sum_k P(X = x_k) \int_{\mathbb{R}} 1_{\{x_k\}}(x) d\mu(x) \\ &= \sum_k P(X = x_k)^2, \end{aligned}$$

where we used Fubini (or MCT) in the penultimate step.

Durrett, 3.4.4. Write

$$\sqrt{S_n} - \sqrt{n} = \frac{S_n - n}{\sqrt{n}} \cdot \frac{1}{\sqrt{S_n/n} + 1},$$

and the second factor converges a.s. to  $1/2$  by SLLN. (You can also directly apply the delta method, by writing  $\sqrt{S_n} - \sqrt{n} = \sqrt{n}(\sqrt{S_n/n} - 1)$ .)

Durrett, 3.4.5. Divide the top and the bottom by  $\sigma\sqrt{n}$  and use SLLN.

1. (a) Omitted. (b) This is the probability that Poisson( $n$ ) random variable is at most its expectation  $n$ , so the limit is  $1/2$  by (a).

2. Immediately we get  $E(X) = 0$ . Let  $\sigma^2 = EX^2 > 0$ . By iterating the distributional equality we get, for any  $n \geq 0$ ,

$$X \stackrel{d}{=} 2^{-n/2}(X_1 + \dots + X_{2^n}),$$

where  $X_i$  are i.i.d. and  $X_i \stackrel{d}{=} X$ . Divide both sides by  $\sigma$ , then send  $n \rightarrow \infty$  and use CLT to get that  $X/\sigma$  is standard Normal.