

Solution to Midterm Exam

1.

(a) Claim: $\log(1-x) \leq -x$ if $0 \leq x < 1$, $\log(1+x) \leq x$ if $0 \leq x < \infty$.

Proof: Let $f(x) = \log(1-x) + x$, $g(x) = \log(1+x) - x$. Then

$$f(0) = 0, \quad f'(x) = \frac{-1}{1-x} + 1 = \frac{-x}{1-x} \leq 0$$

for $0 \leq x < 1$,

$$g(0) = 0, \quad g'(x) = \frac{1}{1+x} - 1 = \frac{-x}{1+x} \leq 0$$

for $0 \leq x < \infty$. Hence

$$0 \leq \left(1 - \frac{x}{n}\right)^n e^{\frac{x}{2}} 1_{(0,n)} = 1_{(0,n)} e^{n \log(1 - \frac{x}{n}) + \frac{x}{2}} \leq e^{-\frac{x}{2}}$$

$$0 \leq \left(1 + \frac{x}{n}\right)^n e^{-2x} 1_{(0,n)} = 1_{(0,n)} e^{n \log(1 + \frac{x}{n}) - 2x} \leq e^{-x}$$

By Dominated Convergence Theorem,

$$\begin{aligned} \lim_{n \rightarrow \infty} \int_0^n \left(1 - \frac{x}{n}\right)^n e^{\frac{x}{2}} dx &= \lim_{n \rightarrow \infty} \int_0^\infty \left(1 - \frac{x}{n}\right)^n e^{\frac{x}{2}} 1_{(0,n)} dx \\ &= \int_0^\infty \lim_{n \rightarrow \infty} \left(1 - \frac{x}{n}\right)^n e^{\frac{x}{2}} 1_{(0,n)} dx \\ &= \int_0^\infty e^{-x/2} dx = 2 \end{aligned}$$

$$\begin{aligned} \lim_{n \rightarrow \infty} \int_0^n \left(1 + \frac{x}{n}\right)^n e^{-2x} dx &= \lim_{n \rightarrow \infty} \int_0^\infty \left(1 + \frac{x}{n}\right)^n e^{-2x} 1_{(0,n)} dx \\ &= \int_0^\infty \lim_{n \rightarrow \infty} \left(1 + \frac{x}{n}\right)^n e^{-2x} 1_{(0,n)} dx \\ &= \int_0^\infty e^{-x} dx = 1 \end{aligned}$$

(b) For any $a > 0$,

$$\int_{(0,a) \times (0,\infty)} |\sin x e^{-xt}| dx dt = \int_{(0,a) \times (0,\infty)} \left| \frac{\sin x}{x} e^{-y} \right| dy dx < \infty$$

By Fubini's Theorem,

$$\begin{aligned}
\int_0^a \frac{\sin x}{x} dx &= \int_0^a \int_0^\infty \sin x e^{-xt} dt dx = \int_0^\infty \int_0^a \sin x e^{-xt} dx dt \\
\int_0^a \sin x e^{-xt} dx &= \int_0^a e^{-xt} d(-\cos x) \\
&= -\cos x e^{-xt} \Big|_0^a + \int_0^a \cos x e^{-xt} (-t) dx \\
&= 1 - \cos a e^{-at} - t \int_0^a e^{-xt} d(\sin x) \\
&= 1 - \cos a e^{-at} - t \sin x e^{-xt} \Big|_0^a + t \int_0^a \sin x e^{-xt} (-t) dx \\
&= 1 - \cos a e^{-at} - t \sin a e^{-at} - t^2 \int_0^a \sin x e^{-xt} dx \\
\implies \int_0^a \sin x e^{-xt} dx &= \frac{1}{t^2 + 1} (1 - \cos a e^{-at} - t \sin a e^{-at}) \\
\implies \lim_{a \rightarrow \infty} \int_0^a \sin x e^{-xt} dx &= \frac{1}{t^2 + 1} \\
\text{and } \left| \int_0^a \sin x e^{-xt} dx \right| &\leq \frac{1}{t^2 + 1} (1 + e^{-at} + t e^{-at}) \in L^1((0, \infty))
\end{aligned}$$

By Dominated Convergence Theorem,

$$\begin{aligned}
\lim_{a \rightarrow \infty} \int_0^a \frac{\sin x}{x} dx &= \lim_{a \rightarrow \infty} \int_0^\infty \int_0^a \sin x e^{-xt} dx dt \\
&= \int_0^\infty \left(\lim_{a \rightarrow \infty} \int_0^a \sin x e^{-xt} dx \right) dt \\
&= \int_0^\infty \frac{1}{t^2 + 1} dt = \frac{\pi}{2}
\end{aligned}$$

(c) For any $t \geq 0$, $0 \leq \log(1 + e^t) - t = \log\left(\frac{1+e^t}{e^t}\right) \leq \ln 2$. Hence for $n > 1$,

$$\left| \frac{1}{n} \log(1 + e^{nf(x)}) \right| \leq \ln 2 + f(x) \in L^1([0, 1]).$$

If $f(x) \geq 0$, $\lim_{n \rightarrow \infty} (1 + e^{nf(x)})^{1/n} = e^{f(x)}$.

If $f(x) < 0$, $\lim_{n \rightarrow \infty} (1 + e^{nf(x)})^{1/n} = 1$.

Hence, $\lim_{n \rightarrow \infty} (1 + e^{nf(x)})^{1/n} = f(x)1_{\{f>0\}}$.
 By Dominated Convergence Theorem,

$$\lim_{n \rightarrow \infty} \frac{1}{n} \int \log(1 + e^{nf(x)}) dx = \int_0^1 \lim_{n \rightarrow \infty} \frac{1}{n} \log(1 + e^{nf(x)}) dx = \int f(x)1_{\{f>0\}} dx.$$

2.

(a) From the translation invariance of λ ,

$$\begin{aligned} \int_{\mathbb{R}^n \times \mathbb{R}^n} |f(x-t)g(t)| \lambda(dxdt) &= \int_{\mathbb{R}^n \times \mathbb{R}^n} |f(y)g(t)| \lambda(dydt) \\ &= \|f\|_1 \|g\|_1 < \infty \end{aligned}$$

By Fubini's Theorem,

$$\int_{\mathbb{R}^n} \left(\int_{\mathbb{R}^n} |f(x-t)g(t)| dt \right) dx = \int_{\mathbb{R}^n \times \mathbb{R}^n} |f(x-t)g(t)| \lambda(dxdt) < \infty$$

Therefore, $(\int_{\mathbb{R}^n} |f(x-t)g(t)| dt) < \infty$, namely, $F_x \in L^1(\mathbb{R}^n)$, for almost every $x \in \mathbb{R}^n$.

(b) By (a),

$$\int_{\mathbb{R}^n} \left| \int_{\mathbb{R}^n} f(x-t)g(t) dt \right| dx \leq \int_{\mathbb{R}^n} \int_{\mathbb{R}^n} |f(x-t)g(t)| dt dx = \|f\|_1 \|g\|_1$$

Thus $f \star g \in L^1(\mathbb{R}^n)$ and $\|f \star g\|_1 \leq \|f\|_1 \|g\|_1$.

(c)

$$f \star g(x) = \int f(x-t)g(t) dt = \int f(y)g(x-y) dy = g \star f(x)$$

$$\int_{\mathbb{R}^n \times \mathbb{R}^n \times \mathbb{R}^n} |f(x-s-t)g(t)h(s)| \lambda(dsdt dx) = \|f\|_1 \|g\|_1 \|h\|_1 < \infty$$

So for almost every $x \in \mathbb{R}^n$,

$$\int_{\mathbb{R}^n \times \mathbb{R}^n} |f(x-s-t)g(t)h(s)| \lambda(dt dx) < \infty.$$

By Fubini's Theorem, for almost every $x \in \mathbb{R}^n$,

$$\begin{aligned}
(f \star g) \star h(x) &= \int \left(\int f(x-s-t)g(t)dt \right) h(s)ds \\
&= \int \left(\int f(x-y)g(y-s)dy \right) h(s)ds \quad \text{by letting } y = t-s \\
&= \int f(x-y) \left(\int g(y-s)h(s)ds \right) dy \\
&= f \star (g \star h)(x) \\
f \star (g+h)(x) &= \int f(x-t)(g(t)+h(t))dt \\
&= \int f(x-t)g(t)dt + \int f(x-t)h(t)dt \\
&= f \star g(x) + f \star h(x)
\end{aligned}$$

Hence \star is commutative, associative and distributive.

(d) There exist some constants $M_j > 0$ and $M > 0$ such that

$$|\partial_{x_j} f| < M_j, \quad |f(x) - f(y)| < M|x - y| \text{ for all } x, y \in \mathbb{R}^n$$

By (a), there exist some x_0 such that $F(x_0, t) = f(x_0 - t)g(t) \in L^1(\mathbb{R}^n)$, so for each x ,

$$|F(x, t) - F(x_0, t)| \leq |f(x-t) - f(x_0-t)||g(t)| < M|x - x_0||g(t)| \in L^1(\mathbb{R}^n)$$

which implies that $F(x, \cdot) \in L^1(\mathbb{R}^n)$ for each x .

Also, $g \in L^1(\mathbb{R}^n)$ implies that $g(t) < \infty$ for almost every t . Thus

$$\partial_{x_j} F(x, t) = (\partial_{x_j} f)(x-t)g(t)$$

exists for almost every t .

Furthermore,

$$|\partial_{x_j}(f(x-t)g(t))| = |(\partial_{x_j} f)(x-t)g(t)| \leq M_j|g(t)| \in L^1(\mathbb{R}^n)$$

By Problem 1(b) from Homework 6, we have

$$\partial_{x_j}(f \star g)(x) = \partial_{x_j} \int f(x-t)g(t)dt = \int (\partial_{x_j} f)(x-t)g(t)dt = (\partial_{x_j} f) \star g(x)$$

3.

(a) Define $f : \mathbb{R}^2 \rightarrow \mathbb{R}$ by $f(x, y) = x + y$. Obviously, f is continuous, hence measurable. So $E_2 = f^{-1}(E) \in \mathcal{B}(\mathbb{R}^2)$.

(b) $\mu \star \nu(\emptyset) = \mu \times \nu(f^{-1}(\emptyset)) = \mu \times \nu(\emptyset) = 0$.
For $A_i \in \mathbb{R}$, which are disjoint, $f^{-1}(A_i) \in \mathcal{B}(\mathbb{R}^2)$ are also disjoint.
Hence

$$\begin{aligned} \mu \star \nu\left(\bigcup A_i\right) &= \mu \times \nu\left(f^{-1}\left(\bigcup A_i\right)\right) = \mu \times \nu\left(\bigcup f^{-1}(A_i)\right) \\ &= \sum \mu \times \nu\left(f^{-1}(A_i)\right) = \sum \mu \star \nu(A_i) \end{aligned}$$

Therefore, $\mu \star \nu$ is a signed measure on \mathbb{R} .

(c) $1_{E_2}(x, t) = 1 \Leftrightarrow x + t \in E \Leftrightarrow x \in E - t \Leftrightarrow 1_{E-t}(x) = 1$.
So $1_{E_2}(x, t) = 1_{E-t}(x)$ and

$$\begin{aligned} \mu \star \nu(E) &= \mu \times \nu(E_2) = \int \int 1_{E_2}(x, t) \mu(dx) \nu(dt) \\ &= \int \int 1_{E-t}(x) \mu(dx) \nu(dt) = \int \mu(E - t) \nu(dt) \end{aligned}$$

When f is a simple function, WLOG, assuming $f = \sum_{i=1}^N a_i 1_{A_i}$ where A_i are disjoint, then

$$\begin{aligned} \int f d(\mu \star \nu) &= \sum_{i=1}^N a_i \mu \star \nu(A_i) = \sum_{i=1}^N \int a_i \mu(A_i - y) \nu(dy) \\ &= \sum_{i=1}^N \int \int f(x + y) 1_{A_i} \mu(dx) \nu(dy) \\ &= \int \int f(x + y) \mu(dx) \nu(dy) \end{aligned}$$

When f is a compactly-supported continuous positive function on \mathbb{R} , and μ, ν are positive measures, $\mu \star \nu$ is also positive. Choose increasing positive simple function sequence h_n such that $h_n \uparrow f$. Then

$$\begin{aligned} \int f d(\mu \star \nu) &= \lim_{n \rightarrow \infty} \int h_n d(\mu \star \nu) \\ &= \lim_{n \rightarrow \infty} \int h_n(x + y) \mu(dx) \nu(dy) \\ &= \int f(x + y) d(\mu \star \nu) \end{aligned}$$

For general compactly supported continuous function f and general signed measure μ, ν , by Hahn decomposition and decomposition for continuous functions

$$\mu = \mu^+ - \mu^-, \quad \nu = \nu^+ - \nu^-, \quad f = f_+ - f_-$$

where μ^+, ν^+ are positive measure, μ^-, ν^- are positive finite measure and f_+, f_- are compactly-supported continuous positive functions. Then we have decomposition for $\mu \star \nu$ and $\mu \times \nu$, that is

$$(\mu \star \nu) = (\mu^+ \star \nu^+ + \mu^- \star \nu^-) - (\mu^+ \star \nu^- + \mu^- \star \nu^+)$$

$$(\mu \times \nu) = (\mu^+ \times \nu^+ + \mu^- \times \nu^-) - (\mu^+ \times \nu^- + \mu^- \times \nu^+)$$

$$\begin{aligned} \int f_+ d(\mu \star \nu) &= \int f_+ d(\mu^+ \star \nu^+) + \int f_+ d(\mu^- \star \nu^-) \\ &\quad - \int f_+ d(\mu^+ \star \nu^-) - \int f_+ d(\mu^- \star \nu^+) \\ &= \int \int f_+(x+y) \mu^+(dx) \nu^+(dy) + \int \int f_+(x+y) \mu^-(dx) \nu^-(dy) \\ &\quad - \int \int f_+(x+y) \mu^+(dx) \nu^-(dy) - \int \int f_+(x+y) \mu^-(dx) \nu^+(dy) \\ &= \int \int f_+(x+y) \mu(dx) \nu(dy) \\ \int f_- d(\mu \star \nu) &= \int \int f_-(x+y) \mu(dx) \nu(dy) \end{aligned}$$

Hence $\int f d(\mu \star \nu) = \int \int f(x+y) \mu(dx) \nu(dy)$.

(d) For $E \in \mathbb{R}$, E_2 is symmetry, namely, $(x, y) \in E_2 \Leftrightarrow (y, x) \in E_2$. Hence,

$$\mu \star \nu(E) = \mu \times \nu(E_2) = \nu \times \mu(E_2) = \nu \star \mu(E)$$

For signed measures $\mu, \nu, \lambda : \mathbb{R} \rightarrow (-\infty, \infty]$, let

$$E_3 = \{(x, s, t) : x + s + t \in E\}$$

$$\begin{aligned}
(\mu \star \nu) \star \lambda(E) &= \int (\mu \star \nu)(E - t) \lambda(dt) \\
&= \int \int \mu(E - t - s) \nu(ds) \lambda(dt) \\
&= \int \int \int 1_{E_3}(x, s, t) \mu(dx) \nu(ds) \lambda(dt) \\
&= \int \int \int 1_{E_3}(x, s, t) \nu(ds) \lambda(dt) \mu(dx) \\
&= (\nu \star \lambda) \star \mu(E) \\
&= \mu \star (\nu \star \lambda)(E) \\
\mu \star (\nu + \lambda)(E) &= \mu \times (\nu + \lambda)(E_2) \\
&= \mu \times \nu(E_2) + \mu \times \lambda(E_2) \\
&= \mu \star \nu(E) + \mu \star \lambda(E)
\end{aligned}$$

So, \star is commutative, associative and distributive.

For Dirac measure δ at 0,

$$\delta \star \mu(E) = \mu \star \delta(E) = \int \mu(E - t) \delta(dt) = \mu(E - 0) = \mu(E)$$

- (e) Lebesgue measure λ is translation invariant. $f, g \in L^1(\mathbb{R}, \lambda)$ implies that

$$\int_{\mathbb{R}^2} |1_E(y) f(y-t) g(t)| \lambda(dy dt) \leq \int_{\mathbb{R}^2} |f(x) g(t)| \lambda(dx dt) = \|f\|_1 \|g\|_1 < \infty$$

Therefore,

$$\begin{aligned}
\mu \star \nu(E) &= \int \mu(E - t) \nu(dt) \\
&= \int \left(\int 1_{E-t}(x) f(x) \lambda(dx) \right) g(t) \lambda(dt) \\
&= \int \int 1_E(y) f(y-t) g(t) \lambda(dy) \lambda(dt) \quad \text{by letting } x = y - t \\
&= \int 1_E(y) \left(\int f(y-t) g(t) \lambda(dt) \right) \lambda(dy) \quad \text{by Fubini's Theorem} \\
&= \int_E f \star g(y) \lambda(dy)
\end{aligned}$$