# UNIVERSITY OF OSLO

## Faculty of mathematics and natural sciences



Please make sure that your copy of the problem set is complete before you attempt to answer anything.

## Problem 1

Discrete time linear filtering.

 $a$  (weight 10p)

Describe the setup for the Kalman-Bucy filter and state the main result (No proof).

### $\mathbf{b}$  (weight 10p)

Let  $X \in \mathbb{R}^n$  and  $Y \in \mathbb{R}^m$  be random vectors. Prove that

 $\widehat{\mathbb{E}}\left[X|\mathcal{L}^Y\right] = \mathbb{E}\left[X\right] + \text{cov}\left(X,Y\right)\text{cov}\left(Y\right)^\oplus\left(Y - \mathbb{E}\left[Y\right]\right).$ 

Hint: You can use without proof the orthogonal projection theorem and you may assume that  $cov(Y)$  is non singular.

c (weight 10p)

Find the recursions for  $\widehat{X}_j$ , the best linear prediction of  $X_j$  given  $\mathcal{L}_j^Y$  :=  $\overline{\text{span}}\left\{1, Y_1, ..., Y_j\right\}$ , and its error  $P_j := \mathbb{E}\left[\left(X_j - \widehat{X}_j\right)^2\right]$  when

$$
X_j = aX_{j-1} + bY_j + \varepsilon_j, \qquad j \ge 1
$$
  

$$
Y_j = cX_{j-1} + \xi_j, \qquad j \ge 1,
$$

where  $\{\varepsilon\}_{j\geq 1}$  and  $\{\xi\}_{j\geq 1}$  are two independent sequences of orthogonal real valued white noises and  $\overline{X}_0 = Y_0 = 0$ .

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## Problem 2

Discrete time nonlinear filtering.

#### $a$  (weight 10p)

Let  $(\Omega, \mathcal{F}, P)$  be a probability space carrying a random variable X and let G be a sub  $\sigma$ -algebra of F. Assume that there exists a regular conditional probability measure  $P(d\omega|X=x)$  on G and it has Radon-Nikodym density  $\rho(\omega, x)$  with respect to a  $\sigma$ -finite measure  $\lambda$  (on  $\mathcal{G}$ ):

$$
P(d\omega|X = x) = \int_B \rho(\omega, x) \lambda(d\omega).
$$

Prove that, then, for every  $\varphi : \mathbb{R} \to \mathbb{R}$ , such that  $\mathbb{E} [|\varphi(X)|] < \infty$  we have that

$$
\mathbb{E}\left[\varphi\left(X\right)|\mathcal{G}\right] = \frac{\int_{\mathbb{R}} \varphi\left(u\right) \rho\left(\omega, u\right) P_X\left(du\right)}{\int_{\mathbb{R}} \rho\left(\omega, u\right) P_X\left(du\right)},
$$

where  $P_X$  is the law of X.

#### **b** (weight  $10p$ )

Describe the setup for finding the nonlinear filter via the Bayes formula and state the main result.

#### $c$  (weight 10p)

Let  $X_j = \varepsilon_j/\varphi(X_{j-1})$  and  $Y_j = X_j + \xi_j$  where  $\varphi$  is a function bounded away from 0  $(\varphi(x) \geq c > 0)$ . The sequences  $(\varepsilon_j)_{j \geq 1}$  and  $(\xi_j)_{j \geq 1}$  are independent of each other and i.i.d.. Moreover,  $\varepsilon_1$  and  $\xi_1$  have densities p and  $\tilde{q}$  with respect to the Lebesgue measure (and enough integrability of its moments). Find a recursion for the random measure  $\pi_i(dx)$  satisfying

$$
\int_{\mathbb{R}} f(x) \pi_j(dx) = \mathbb{E} \left[ f(X_j) | \mathcal{F}_j^Y \right].
$$

## Problem 3

Continuous time linear filtering.

Suppose we want to estimate the value of a constant parameter  $\theta$ , based on observations  $Y_t$  satisfying

$$
dY_t = \theta M_t dt + N_t dB_t,
$$

where  $M_t$  and  $N_t$  are known deterministic functions and B is a Brownian motion.

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## $a$  (weight 10p)

Find the equations for  $\hat{\theta}$ , the best linear estimator of  $\theta$  using the values of the process Y up to time t, and  $P_t = \mathbb{E}\left[\left(\theta - \widehat{\theta}_t\right)^2\right].$ 

## $\mathbf{b}$  (weight 10p)

Check that  $P_t = \left(P_0^{-1} + \int_0^t M_s^2 N_s^{-2} ds\right)^{-1}$  solves the equation for  $P_t$  and prove that

$$
\widehat{\theta}_t = \frac{\widehat{\theta}_0 P_0^{-1} + \int_0^t M_s N_s^{-2} dY_s}{P_0^{-1} + \int_0^t M_s^2 N_s^{-2} ds}.
$$

## Problem 4

Continuous time nonlinear filtering.

a (weight 10p)

Describe the general setup for the nonlinear filtering problem in continuous time and the main result (No proof).

 $\mathbf{b}$  (weight 10p)

State and sketch the proof of Zakai equation. By sketching I mean that there is no need to justify some applications of dominated convergence in the proof.