

CH 10 BAYESIAN ESTIMATION

SO FAR, IAV MLE: MAX PDF OF Y
GIVEN θ

IN MAXIMUM A POSTERIORI (MAP) ESTIMATION,
(A BAYESIAN APPROACH)

IN MAP: MAX. PDF OF θ
GIVEN Y

USING "BAYES FORMULA" WE GET:

$$P(\theta | Y) = \frac{P(Y | \theta) P(\theta)}{P(Y)}$$

$P(A|B) \Leftrightarrow$ PROB. OF EVENT A GIVEN B

$$f_{\text{MAP}}(\theta | y) = \frac{\boxed{f_{\text{ML}}(y | \theta)} \cdot \boxed{g(\theta)}}{f_y(y)}$$

$f_{\text{ML}}(y | \theta)$ LIKELIHOOD FCN FROM MLE

$g(\theta)$ A PRIORI KNOWLEDGE

$f_y(y)$ PDF TO GET MEASUREMENT y
(INDEP. OF θ)

$$\begin{aligned} \hat{\theta}_{\text{MAP}}(y) &= \arg \max_{\theta} f_{\text{MAP}}(\theta | y) \\ &= \arg \min_{\theta} (-\log f_{\text{MAP}}(\theta | y)) \\ &= \arg \min_{\theta} \left\{ \underbrace{(-\log f_{\text{ML}}(y | \theta))} + \underbrace{(-\log g(\theta))} - \underbrace{(-\log f_y(y))} \right\} \end{aligned}$$

EXAMPLE :

A PRIORI : $\Theta \sim \mathcal{N}(\bar{\Theta}, \Sigma_{\Theta})$

$$\boxed{g(\Theta)} = \text{const.} \cdot \exp\left(-\frac{1}{2}(\Theta - \bar{\Theta})^T \Sigma_{\Theta}^{-1}(\Theta - \bar{\Theta})\right)$$

$$-\log g(\Theta) = \frac{1}{2}(\Theta - \bar{\Theta})^T \Sigma_{\Theta}^{-1}(\Theta - \bar{\Theta})$$

REGULARIZATION TERM $-\log g(\Theta)$ IS
ADDED TO ML-OBJECTIVE, IT CAN INTRODUCE
BIAS, CAN INTERPRET AS

"PSEUDO MEASUREMENT" OF Θ
WITH VALUE $\bar{\Theta}$ AND ERROR Σ_{Θ}

EX 2 RE INTERPRETATION OF "TRAJECTORY FITTING"

$$y(k) = a(1) \cdot y(k-1) + a(2) y(k-2) + b(1)u(k-1) + \varepsilon_e(k)$$

MEASUREMENTS : $y^{\text{MEAS}}(k) = y(k) + \varepsilon_y(k)$
 $u^{\text{MEAS}}(k) = u(k) + \varepsilon_u(k)$

ASSUMED $\varepsilon_e(k) \sim \mathcal{N}(0, G_e^2)$

$$\varepsilon_y(k) \sim \mathcal{N}(0, G_y^2)$$

$$\varepsilon_u(k) \sim \mathcal{N}(0, G_u^2)$$

$$\Theta = \begin{bmatrix} a(1) \\ a(2) \\ b(1) \\ u(1) \\ \vdots \\ u(N) \\ y(1) \\ \vdots \\ y(N) \end{bmatrix}$$

WE FOUND $\hat{\Theta}$ BY

$$\hat{\Theta} = \underset{\Theta}{\text{argmin}} \sum \left\{ \left(\frac{y^{\text{MEAS}} - y}{G_y} \right)^2 + \left(\frac{u^{\text{MEAS}} - u}{G_u} \right)^2 \right.$$

A PRIORI
KNOWLEDGE

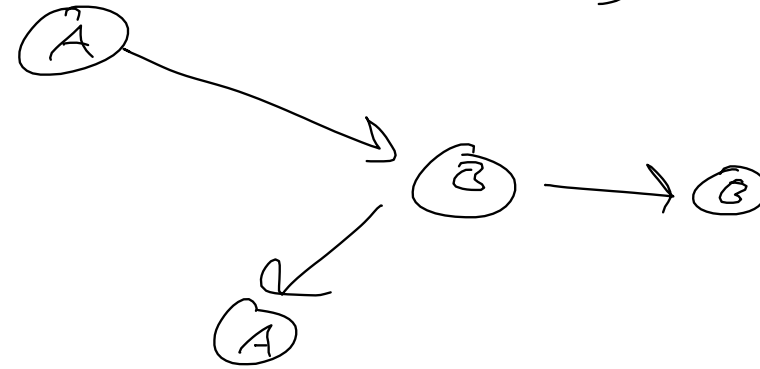
$$+ \left(\frac{y(k) - a(1) \cdot y(k-1) \dots}{G_e} \right)^2$$

2
4

BRAINSTORM FOR EXAMPLES

A • HEATING WATER ON A PLATE 7

B • COLLIDING BALLS
(ELASTICITY, ...) 5



C • CRANE

(INPUT: SPEED OF MOTOR,
OUTPUT: MASS)

(LENGTH OF ROPE) 4

D • HOT AIR BALLOON 10

INPUT: HEATING (MEASURE)

OUTPUT: ALTITUDE (")

MODELLING



E • ANIMAL POPULATION

MONKEYS
&
SNAKES 6

EXAMPLE: HOT AIR BALLOON

INPUTS: HEATING u

OUTPUTS: ALTITUDE y

CHOICES TO MAKE

- BLACK OR WHITE BOX?
- CONT. VS. DISCR.

DECISIONS

- ONLY VERTICAL MOTION
- OUTSIDE TEMP. CONST.
- PRESSURE DEP. ON ALTITUDE
- NEGLECT COOLING VIA INCOMING AIR WHEN BALLOON COOLS
- NEGLECT VARYING HEAT CAPACITY
- NEGLECT AIR FRICTION

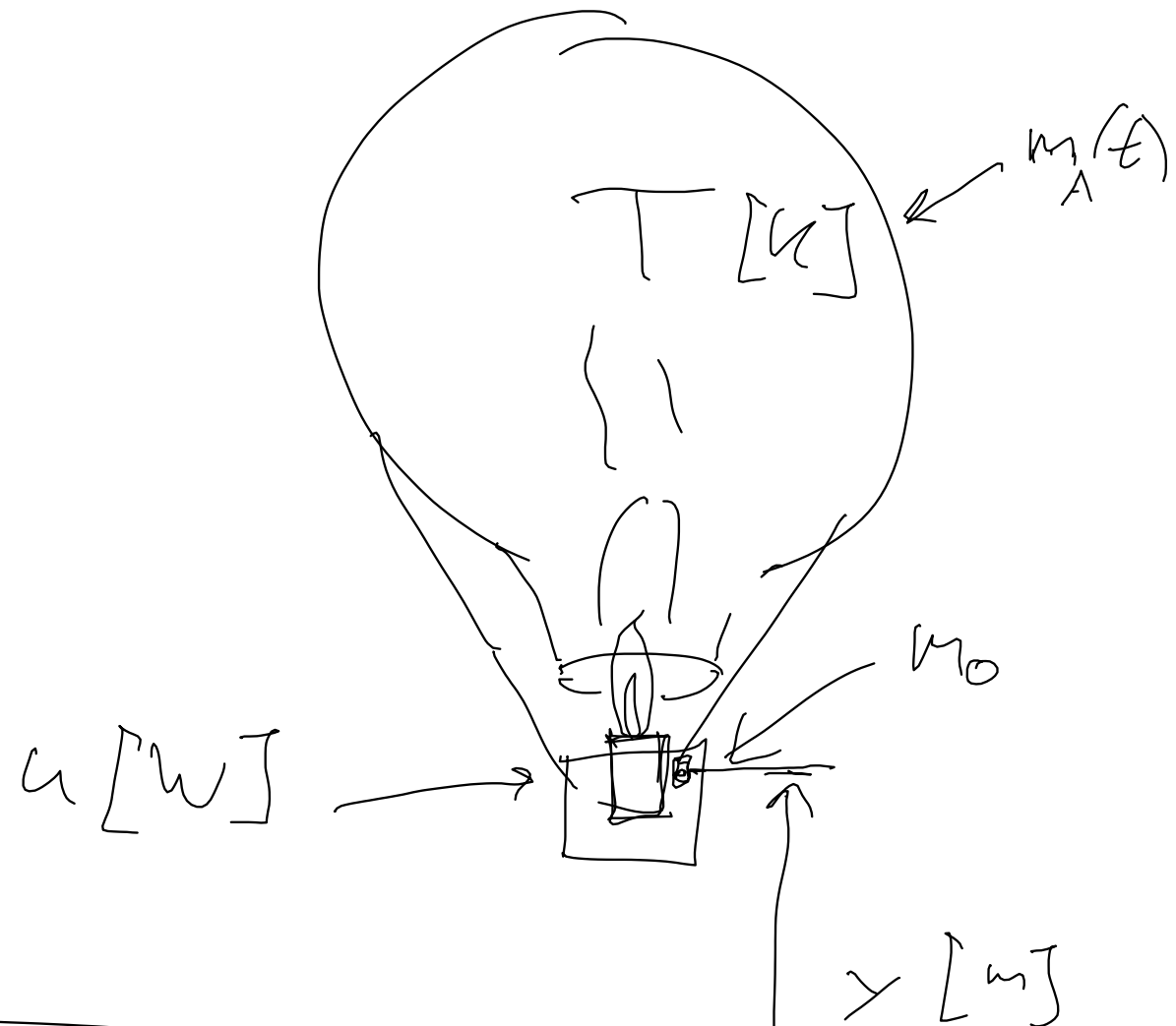
$$m \ddot{y} = -mg + F_{\text{BUOYANCY}}$$

PARAMETERS

- m MASS OF WHOLE APPARATUS
- $\dot{T}$ TEMPERATURE IN BALLOON

$$F_{\text{BUOYANCY}} = ?$$

$$m \sim T$$
$$\dot{T} = ?$$





$$[pV = n k_B T]$$

ode45

$$p(y)V = T(t) \cdot m_A(t) \cdot \frac{k}{T(t)}$$

$$m_A(t) = \left[\frac{V}{k} \right] \frac{p(y)}{T(t)}$$

$p(y)$: known

$$F_{\text{BUOYANCY}}(y) = g \cdot m_{\text{OUTSIDE-AIR}}(y)$$

$$m_{\text{OUTSIDE-AIR}}(y) = \left[\frac{V}{k} \right] \frac{p(y)}{T_0}$$

$$m(t) \quad \boxed{m(t) = m_0 + m_A(t) = m_0 + \left[\frac{V}{k} \right] \frac{p(y(t))}{T(t)}}$$

$$\text{STATE: } \begin{bmatrix} y \\ \dot{y} \\ T \end{bmatrix} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$$

$$\boxed{C \cdot \dot{T} = u - k_2(T - T_0)}$$

$$\begin{cases} \dot{y} = \dot{y} \\ \ddot{y} = -g + \frac{F_B(y)}{m(t)} \\ \dot{T} = \frac{u}{C} - \frac{k_2}{C} T + \frac{k_2}{C} T_0 \end{cases} \quad \begin{cases} \dot{x}_1 \equiv x_2 \\ \dot{x}_2 = -g + \frac{F_B(x_1)}{m(t)} \end{cases}$$

$$\boxed{\dot{x}_3 = \frac{u}{C} - \frac{k_2}{C} x_3 + \frac{k_2}{C} T_0}$$

$$\begin{cases} \dot{x} = f(x, u) \\ y = g(x, u) \end{cases}$$