

**BAYESIAN INVERSE PROBLEMS FOR
BURGERS AND HAMILTON-JACOBI
EQUATIONS WITH WHITE NOISE FORCING**

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BURGERS AND HAMILTON-JACOBI EQUATIONS

We consider Burgers equation with white noise forcing:

$$\frac{\partial u}{\partial t} + (u \cdot \nabla)u = f(x)\dot{W}(t), \quad u \in \mathbb{R}^d, \quad x \in \mathbb{R}^d.$$

Assume: $f(x) = -\nabla F(x)$, and $u(t, x) = \nabla \phi(t, x)$;

ϕ satisfies the Hamilton-Jacobi equation:

$$\frac{\partial \phi(x, t)}{\partial t} + \frac{1}{2} |\nabla \phi(x, t)|^2 + F(x)\dot{W}(t) = 0.$$

- Burgers equation is a model for studying turbulence. It also has applications in non-equilibrium mechanics.

- We are interested in long time behavior: we consider the equations on $(-\infty, T]$.

- Suppose at t_1, t_2, \dots, t_m , observations are made for velocity u and velocity potential ϕ (subject to Gaussian noise);

we make inference on the white noise forcing on $(-\infty, T]$.

- Bayesian inverse problem for Navier-Stokes equations with model errors (stochastic forcing) is considered by Cotter, Dashti, Robinson and Stuart (2009)

SOLUTION FORMULA

- Given an initial condition $\phi(x, t_0) = \phi_0(x)$, $\phi(\cdot, t)$ is determined by *Lax operator*:

$$\phi(\cdot, t) = \mathcal{K}_{t_0, t}^W \phi_0.$$

- Lax-Oleinik formula:

$$\phi(x, t) = \inf \left\{ \phi_0(\gamma(t_0)) + \int_{t_0}^t \frac{1}{2} |\dot{\gamma}(\tau)|^2 - F(\gamma(\tau)) \dot{W}(\tau) d\tau \right\},$$

where inf is taken with respect to all *absolutely continuous curves* γ s.t. $\gamma(t) = x$.

- We are interested in solutions that exist for all time, i.e.

$$\phi(\cdot, t) = \mathcal{K}_{t_0, t}^W \phi(\cdot, t_0), \quad \forall t_0 < t.$$

PERIODIC SETTING

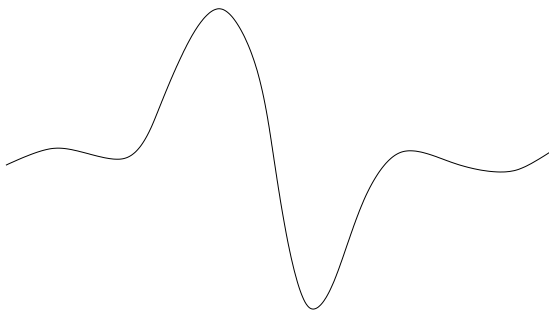
- E, Khanin, Mazel and Sinai (2000) and Iturriaga and Khanin (2003):

there is a unique solution ϕ (within an *additive constant*) that exists for all time, i.e.

$$\phi(\cdot, t) = \mathcal{K}_{t_0, t}^W \phi(\cdot, t_0), \quad \forall t_0 < t.$$

- For all t : $\phi(\cdot, t)$ is continuous, and Lipschitz.
- There is a unique spatially periodic solution $u(t, x)$ for the Burgers equation that exists for all time.

NON-PERIODIC SETTING



- For potential $F(x)$ with a “big” maximum and a “big” minimum, H. and Khanin (2003) show that there is a solution ϕ and a solution u that exist for all time.
- They are limit of finite time solutions with zero initial conditions.

BAYESIAN INVERSE PROBLEM FOR H-J EQUATION

- Formulation: As ϕ is uniquely determined within a constant, and is continuous

$$\mathcal{G}(W) = \{\phi^W(x_i, t_i) - \phi^W(x_0, t_0), i = 1, \dots, m\} \in \mathbb{R}^m,$$

is uniquely determined by W .

Let y be a noisy observation of $\mathcal{G}(W)$:

$$y = \mathcal{G}(W) + \sigma.$$

The prior probability μ_0 is the Wiener measure on $C(-\infty, t_{\max}]$ ($t_{\max} = \max t_i$).

Determine $\mu^y(W) = \mathbb{P}(W|y)$.

BAYESIAN INVERSE PROBLEM FOR H-J EQUATION

Assuming a Gaussian noise $\sigma \sim \mathcal{N}(0, \Sigma)$, we aim to show:

- Bayes' formula holds:

$$\frac{d\mu^y}{d\mu_0} \propto \exp(-\Phi(W; y))$$

where

$$\Phi(W; y) = \frac{1}{2} |y - \mathcal{G}(W)|_{\Sigma}^2 = \frac{1}{2} \langle \Sigma^{-1/2}(y - \mathcal{G}(W)), \Sigma^{-1/2}(y - \mathcal{G}(W)) \rangle.$$

- The posterior μ^y is well-posed; in particular

$$d_{\text{Hell}}(\mu^y, \mu^{y'}) \leq c(r) |y - y'|_{\mathbb{R}^m},$$

when $|y|_{\mathbb{R}^m} \leq r$ and $|y'|_{\mathbb{R}^m} \leq r$.

BANACH SPACE SETTING

For $y = \mathcal{G}(x) + \sigma$, $x \in X$ a **Banach space**:

Cotter, Dashti, Robinson and Stuart showed:

(I) If $\mathcal{G} : X \rightarrow \mathbb{R}^m$ is measurable, e.g. when it is continuous with respect to x , the Bayes' formula holds.

(II) When μ_0 is Gaussian,
when $|y_1|_{\mathbb{R}^m} \leq r$, $|y_2|_{\mathbb{R}^m} \leq r$

$$|\Phi(x; y_1) - \Phi(x; y_2)| \leq K(r)(1 + \|x\|_X^q)|y_1 - y_2|_{\mathbb{R}^m}$$

then the posterior measure μ^y is well-posed, i.e.

$$d_{\text{Hell}}(\mu^{y_1}, \mu^{y_2}) \leq c(r)|y_1 - y_2|_{\mathbb{R}^m}.$$

METRIC SPACE SETTING

- Our space $C(-\infty, t_{\max}]$ is not Banach;
- It is a metric space with the metric:

$$D(W_1, W_2) = \sum_{n=1}^{\infty} \frac{1}{2^n} \frac{\sup_{-n \leq t \leq t_{\max}} |W_1(t) - W_2(t)|}{1 + \sup_{-n \leq t \leq t_{\max}} |W_1(t) - W_2(t)|}.$$

- We need to formulate Bayesian inverse problems for **metric** spaces.
- For a metric space X , condition (I) of Cotter et al. still holds: If \mathcal{G} is continuous, then the Bayes' formula holds.

METRIC SPACE SETTING

For well-posedness: Condition (II) needs to be generalized.

(III)

i) Φ is locally bounded: for $r > 0$, if $|y|_{\mathbb{R}^m} \leq r$

$$0 \leq \Phi(x; y) \leq M(r),$$

for $x \in X(r) \subset X$, $\mu_0(X(r)) > 0$.

ii) There is a $G : \mathbb{R} \times X \rightarrow \mathbb{R}$: $G(r, \cdot) \in L^2(X, d\mu_0)$, and

$$|\Phi(x; y) - \Phi(x; y')| \leq G(r, x)|y - y'|_{\mathbb{R}^m},$$

when $|y|_{\mathbb{R}^m} \leq r$ and $|y'|_{\mathbb{R}^m} \leq r$.

Then

$$d_{\text{Hell}}(\mu^y, \mu^{y'}) \leq c(r)|y - y'|_{\mathbb{R}^m}.$$

PERIODIC H-J EQUATION

- We consider the **periodic case** first:

the forcing function $f(x)$ and forcing potential $F(x)$ are periodic; problems are on \mathbb{T}^d .

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$$\mathcal{G}(W) = \{\phi(x_i, t_i) - \phi(x_0, t_0) : i = 1, \dots, m\}.$$

- To show the validity of the Bayes' formula, we show that $\mathcal{G} : C(-\infty, t_{\max}] \rightarrow \mathbb{R}^m$ is continuous.
- To show well-posedness, we show conditions (III)(i) and (III)(ii).

PERIODIC H-J EQUATION

- First we show that \mathcal{G} is continuous.
- From the Lax-Oleinik formula: when $D(W_k, W) \rightarrow 0$, there are c_k independent of x_i and t_i s.t.

$$\phi^{W_k}(x_i, t_i) - \phi^W(x_i, t_i) - c_k \rightarrow 0, \quad i = 0, 1, \dots, m.$$

- $\mathcal{G}(W) = \{\phi^W(x_i, t_i) - \phi^W(x_0, t_0), \quad i = 1, \dots, m\}$,

$\mathcal{G} : C(-\infty, t_{\max}] \rightarrow \mathbb{R}^m$ is continuous.

- The Bayes' formula holds:

$$\frac{d\mu^y}{d\mu_0} \propto \exp(-\Phi(W; y)) = \exp(-\frac{1}{2}|y - \mathcal{G}(W)|_{\Sigma}^2).$$

PERIODIC H-J EQUATION

- For well-posedness, we show (III)(i) and (III)(ii):

$$(III)(i) \quad 0 \leq \Phi(W; y) \leq M(r) \quad \text{when } |y|_{\mathbb{R}^m} \leq r;$$

$W \in X(r) \subset C(-\infty, t_{\max}]$, $X(r)$ of positive μ_0 measure.

from the Lax operator

$$\begin{aligned} |\mathcal{G}(W)|_{\mathbb{R}^m} &= |\{\phi^W(x_i, t_i) - \phi^W(x_0, t_0)\}|_{\mathbb{R}^m} \\ &\leq c \left(1 + \sum_{i=1}^m \max_{t_0-1 \leq \tau \leq t_i} |W(\tau) - W(t_i)|^2 \right). \end{aligned}$$

$$\Phi(W; y) = \frac{1}{2} |y - \mathcal{G}(W)|_{\Sigma}^2 \leq c(r + |\mathcal{G}(W)|_{\mathbb{R}^m})^2 \quad \text{when } |y|_{\mathbb{R}^m} \leq r.$$

Fixing $M > 0$, the set W s.t. $|\mathcal{G}(W)|_{\mathbb{R}^m} < M$ has a positive Wiener measure. (III)(i) is thus shown.

(III)(ii)

$$|\Phi(W; y) - \Phi(W; y')| \leq G(r, W)|y - y'|_{\mathbb{R}^m}$$

when $|y|_{\mathbb{R}^m} \leq r$, $|y'|_{\mathbb{R}^m} \leq r$ and $G(r, \cdot)$ is in $L^2(C(-\infty, t_{\max}], d\mu_0)$.

- With $\Phi(W; y) = \frac{1}{2}|y - \mathcal{G}(W)|_{\Sigma}^2$:

$$\begin{aligned} |\Phi(W; y) - \Phi(W; y')| &\leq \frac{1}{2} \|\Sigma^{-1/2}\|_{\mathbb{R}^m, \mathbb{R}^m}^2 (|y|_{\mathbb{R}^m} + |y'|_{\mathbb{R}^m} + 2|\mathcal{G}(W)|_{\mathbb{R}^m}) |y - y'|_{\mathbb{R}^m} \\ &\leq \|\Sigma^{-1/2}\|_{\mathbb{R}^m, \mathbb{R}^m}^2 (r + |\mathcal{G}(W)|_{\mathbb{R}^m}) |y - y'|_{\mathbb{R}^m}. \end{aligned}$$

With the bound in the previous slice, this is square integrable.

NONPERIODIC H-J EQN

- We consider the **non-periodic** case where the forcing potential $F(x)$ has a “big” maximum and a “big” minimum.
- To show Bayes' formula, we show that

$$\mathcal{G}(W) = \{\phi^W(x_i, t_i) - \phi^W(x_0, t_0), i = 1, \dots, m\}$$

is continuous from $C(-\infty, t_{\max}]$ to \mathbb{R}^m .

This is shown similarly as in the periodic case: there are constants c_k so that:

$$\lim_{k \rightarrow \infty} \phi^{W_k}(x_i, t_i) - \phi^W(x_i, t_i) - c_k = 0.$$

- The Bayes' formula thus holds.

NONPERIODIC H-J EQN

- The well-posedness of the posterior measure μ^y , we show conditions (III)(i) and (III)(ii).
- We show the bound:

$$|\mathcal{G}(W)|_{\mathbb{R}^m} \leq S(W),$$

where

$$S(W) = c + c \sum_{i=1}^m \sum_{l=T_i(W)}^{t_{\max}} (1 + \max_{l \leq \tau \leq l+1} |W(\tau) - W(l+1)|^2).$$

- The constant $T_i(W)$ depends on the Wiener path W .

NONPERIODIC H-J EQN

- The Lax operator

$$\phi(\cdot, t) = \mathcal{K}_{s,t}^W \phi(\cdot, s);$$

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$$\phi(x_i, t_i) = \inf_{\gamma, \gamma(t_i)=x_i} \left\{ \phi(\gamma(s), s) + \int_{t_0}^t \frac{1}{2} |\dot{\gamma}(\tau)|^2 - F(\gamma(\tau)) \dot{W}(\tau) d\tau \right\}.$$

- We show that all the minimizers γ are inside a compact set at a time;

γ must be inside the compact set at a time larger than $T_i(W)$, which is independent of s .

- For the condition (III)(i): we show

$$\Phi(W; y) = \frac{1}{2} |y - \mathcal{G}(W)|_{\Sigma}^2 \leq c(r + |\mathcal{G}(W)|_{\mathbb{R}^m})^2 \leq c(r + S(W))^2$$

is less than $M(r)$ for $W \in X(r)$ of positive Wiener measure.

- There is a constant T such that the set of paths W with $T_i(W) > T$ has a positive Wiener measure.
- We can choose a constant M s.t. out of these paths, the set of paths W such that $S(W) < M$ has a positive Wiener measure.

- For the condition (III)(ii):

$$\begin{aligned} |\Phi(W; y) - \Phi(W; y')| &\leq \|\Sigma^{-1/2}\|_{\mathbb{R}^m, \mathbb{R}^m}^2 (r + G(W)) |y - y'|_{\mathbb{R}^m}. \\ &\leq \|\Sigma^{-1/2}\|_{\mathbb{R}^m, \mathbb{R}^m}^2 (r + S(W)) |y - y'|_{\mathbb{R}^m}. \end{aligned}$$

- To show that $G(r, W) = \|\Sigma^{-1/2}\|_{\mathbb{R}^m, \mathbb{R}^m}^2 (r + S(W))$ is in $L^2(C(-\infty, t_{\max}], d\mu_0)$

we show $S(W) \in L^2(C(-\infty, t_{\max}], d\mu_0)$.

- This is achieved by using estimates for the convergence rates for the law of large numbers.

- There are shocks where the solution u is discontinuous; u is not defined everywhere, but $u(\cdot, t) \in L^1_{\text{loc}}(\mathbb{R}^d)$ for all t .
- For $i = 1, \dots, m$, let $l_i : L^1_{\text{loc}}(\mathbb{R}^d) \rightarrow \mathbb{R}$ be continuous and bounded.
- Define

$$\mathcal{G}(W) = (l_1(u(\cdot, t_1)), \dots, l_m(u(\cdot, t_m))) \in \mathbb{R}^m.$$

- Noisy observation

$$y = \mathcal{G}(W) + \sigma.$$

- Determine $\mu^y(W) = \mathbb{P}(W|y)$.