# Two Results Concerning Asymptotic Behavior of Solutions of the Burgers Equation with Force 

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#### Abstract

We consider the Burgers equation with an external force. For the case of the force periodic in space and time we prove the existence of a solution periodic in space and time which is the limit of a wide class of solutions as $t \rightarrow \infty$. If the force is the product of a periodic function of $x$ and white noise in time, we prove the existence of an invariant distribution concentrated on the space of spaceperiodic functions which is the limit of a wide class of distributions as $t \rightarrow \infty$.


[^0]
## 1. FORMULATION OF THE RESULTS

We consider the one-dimensional Burgers equation with force having the form

$$
\begin{equation*}
u_{t}+u \cdot u_{x}=\mu u_{x x}+F^{\prime}(x) B(t), \quad-\infty<x<\infty \tag{1}
\end{equation*}
$$

Here $F(x)$ is a $C^{1}$-periodic function of period $x_{0}$. The assumptions concerning $B$ will be formulated later. The initial data $u(x ; 0)$ are derivatives $u(x ; 0)=v^{\prime}(x)$, where $v(x)$ are typical realizations of some random process. The probability distribution corresponding to $v$ is denoted by $P_{0}$. It is defined on the natural $\sigma$-algebra of subsets of the space $V$ of absolutely continuous functions $v(x)$. We assume that:

1. There exists a constant $C_{0}$ such that with $P_{0}$-probability 1

$$
|v(x)| \leqslant C_{0}
$$

for all $x$.

[^1]2. $A=E\{\exp [-v(x)]\}$ does not depend on $x$.
3. There exists $\gamma, 0<\gamma<1 / 2$, such that for $P_{0}$-almost every $v(x)$,
$$
\lim _{n \rightarrow \infty} \sup _{\substack{a \in\left[0, x_{0}\right] \\|m| \leqslant n, m \in \mathbb{Z}^{1}}} \frac{1}{2\left[n^{\nu}\right]+1}\left|\sum_{|k-m| \leqslant\left[n^{\nu}\right]} e^{-v\left(a+k x_{0}\right)}-A\right|=0
$$

It is easy to give concrete examples of $P_{0}$ for which condition 3 is true.
Theorem 1. Let $B$ be a continuous periodic function of period $\tau_{0}$. There exists a solution $u^{(0)}(x, t)$ of (1) periodic in $x$ with period $x_{0}$ and periodic in time with period $\tau_{0}$ such that for $P_{0}$-a.e. $v$

$$
\lim _{t \rightarrow \infty}\left[u(x, t)-u^{(0)}(x, t)\right]=0
$$

for any $x,-\infty<x<\infty$.
Remarks. 1. Our method of proof also gives an explicit expression for $u^{(0)}(x, t)$.
2. The theorem remains true if the force in the Burgers equation takes the form $\partial F(x, t) / d x$, where $F(x, t)$ is a function periodic in space with period $x_{0}$ and periodic in time with period $\tau_{0}$.
3. The theorem remains true for bounded functions $v$ such that $v^{\prime}(x) \rightarrow 0$ as $x \rightarrow \pm \infty, v(x) \rightarrow$ const as $x \rightarrow \pm \infty$.
4. The convergence in Theorem 1 is pointwise. After giving the proof, we discuss stronger statements concerning the character of convergence.

In Theorem 2 we assume that $B(t)$ is a white noise. This means that for any $t_{1}, t_{2}, t_{1}<t_{2}$, a random variable $b\left(t_{1}, t_{2}\right)=\int_{t_{1}}^{t_{2}} B(\tau) d \tau$ is defined such that:
$\left(\mathrm{a}_{1}\right) \quad b\left(t_{1}, t_{2}\right)$ has the Gaussian distribution with mean value equal to zero and dispersion $E b^{2}\left(t_{1}, t_{2}\right)=\sigma\left(t_{2}-t_{1}\right)$ for some $\sigma>0$.
$\left(\mathrm{a}_{2}\right)$ For nonoverlapping intervals $\left(t_{1}^{\prime}, t_{2}^{\prime}\right)$ and $\left(t_{1}^{\prime \prime}, t_{2}^{\prime \prime}\right)$ the random variables $b\left(t_{1}^{\prime}, t_{2}^{\prime}\right)$ and $b\left(t_{1}^{\prime \prime}, t_{2}^{\prime \prime}\right)$ are independent.

Denote by $M\left(\left(t_{1}, t_{2}\right)\right.$ the least $\sigma$-algebra generated by all $b\left(t_{1}^{\prime}, t_{2}^{\prime}\right)$, where $t_{1}<t_{1}^{\prime}<t_{2}^{\prime}<t_{2}$, and let $\left\{T^{\prime}\right\}$ be the measure-preserving flow on the space of all random variables $b\left(t_{1}^{\prime}, t_{2}^{\prime}\right)$, where each $T^{t}$ transforms $M\left(t_{1}, t_{2}\right)$ to $M\left(t_{1}-t, t_{2}-t\right)$ and

$$
\left(T^{t} b\right)\left(t_{1}^{\prime}, t_{2}^{\prime}\right)=b\left(t_{1}-t^{\prime}, t_{2}-t^{\prime}\right)
$$

for any $t_{1}^{\prime}, t_{2}^{\prime}, t_{1}+t<t_{1}^{\prime}<t_{2}^{\prime}<t_{2}+t$.
Assume that $P_{0}$ satisfies the same conditions as in Theorem 1.

Theorem 2. Let $P_{t}$ be the natural probability distribution on the space of solutions $u(x, t)$ of (1) induced by $P_{0}, 0<t<\infty$. Then $P_{t}$ converges weakly as $t \rightarrow \infty$ to some probability distribution $Q$ which does not depend on $P_{0}$ and is concentrated on the space of functions periodic in $x$ with period $x_{0}$.

The proof of Theorem 1 is given in Section 2. In Section 3 we expound the proof of Theorem 2. The actual statement which we show is the following. For any $\varepsilon<0$ we find $t_{0}(\varepsilon)$, a set $C \in M\left(0, t_{0}(\varepsilon)\right)$, and a functional $H_{x}\left(\left\{B\left(t_{1}, t_{2}\right)\right\}, 0 \leqslant t_{1}, t_{2} \leqslant t_{0}(\varepsilon)\right)$ defined on $C$ and such that $\operatorname{Prob}(C) \geqslant 1-\varepsilon$ and if $T^{-t} b \in C$, then

$$
\left|u(x, t)-H_{x}\left(b\left(t_{1}, t_{2}\right), t-t_{0}(\varepsilon) \leqslant t_{1}, t_{2} \leqslant t\right)\right| \leqslant \varepsilon
$$

In other words, for increasing $t$, the solution $u(x, t)$ becomes a functional of the realization of white noise $B(\tau), 0<\tau<t$, with "short memory." This memory can be estimated in a more precise way. The functional $H_{x}$ depends periodically on $x$. Theorems 1 and 2 are valid also for the multidimensional Burgers equation. Only small modifications in the proofs are needed.

## 2. PROOF OF THEOREM 1

After the appropriate rescaling of $x$ and $\mu$ we may assume that the period $\tau_{0}=1$. We use the Cole-Hopf substitution $u=-2 \mu\left(\varphi_{x} / \varphi\right),{ }^{(1,2)}$ and get for $\varphi$ the equation

$$
\begin{equation*}
\varphi_{t}=\mu \varphi_{x x}-\frac{1}{2 \mu} F(x) B(t) \varphi \tag{2}
\end{equation*}
$$

The Feynman-Kac formula ${ }^{(3)}$ makes it possible to write down $\varphi$ as a functional integral. Namely, denote by $\Pi_{W_{1}, W_{2}}^{\left(t_{1}, t_{2}\right)}$ the corresponding Wiener measure on the space of continuous functions $w(\tau), t_{2} \leqslant \tau \leqslant t_{1}$, such that $w\left(t_{1}\right)=w_{1}, w\left(t_{2}\right)=w_{2}$. Then

$$
\begin{align*}
\varphi(x ; t)= & \int_{-\infty}^{\infty} d y\{\exp [-v(y)]\} \\
& \times \int\left\{\exp \left[\int_{0}^{t} F(W(\tau)) B(\tau) d \tau\right]\right\} d \Pi_{x, y}^{(t, 0)}(W) \tag{3}
\end{align*}
$$

Put $t=t_{0}, t-j=t_{j}, j \geqslant 1$, and $j \in \mathbb{Z}^{1}$, and find $r$ such that $t_{r+2}<$ $0<t_{r+1}$. Fix the numbers $a_{1}, a_{2}, \ldots, a_{r}, a_{r+1}, a_{j} \in\left[0, x_{0}\right.$ ) and rewrite (3) as follows:

$$
\begin{align*}
\varphi(x ; t)= & \int_{0}^{x_{0}} \cdots \int_{0}^{x_{0}} d a_{1} d a_{2} \cdots d a_{r} d a_{r+1} \sum_{\substack{m_{1}, m_{2}, \ldots, m_{r+1} \\
m_{j} \in \mathbb{Z}^{1}}} K_{t_{1}}\left(x, a_{1}+m x_{0}\right) \\
& \times \prod_{j=2}^{r} K_{t_{j}}\left(a_{j-1}+m_{j} x_{0}, a_{j}+m_{j} x_{0}\right) \\
& \times K_{f}\left(a_{r}+m_{r} x_{0}, a_{r+1}+m_{r+1} x_{0}\right) e^{-v\left(a_{r+1}+m_{r+1} x_{0}\right)} \tag{4}
\end{align*}
$$

Here

$$
\begin{aligned}
& K_{t_{j}}\left(W_{1}, W_{2}\right)=\int\left\{\exp \left[\int_{t_{j-1}}^{t_{j}} F(W(\tau)) B(\tau) d t\right]\right\} d \Pi_{\left(W_{1}, W_{2}\right)}^{\left(t_{1}, t_{j}\right)}(W) \\
& K_{f}\left(W_{1}, W_{2}\right)=\int\left\{\exp \left[\int_{0}^{t_{r}} F(W(\tau)) B(\tau) d t\right]\right\} d \Pi_{\left(W_{1}, W_{2}\right)}^{\left(t_{1}, 0\right)}(W)
\end{aligned}
$$

The periodicity of $F$ in space and that of $B$ in time imply the following relations:

$$
\text { 1. } \quad K_{t_{j}}\left(W_{1}, W_{2}\right)=K_{t_{j}}\left(W_{1}+m x_{0}, W_{2}+m x_{0}\right), \quad 2 \leqslant j \leqslant r
$$

for all $m \in \mathbb{Z}^{1}$.

$$
\text { 2. } \quad K_{t_{2}}\left(W_{1}, W_{2}\right)=K_{t_{3}}\left(W_{1}, W_{2}\right)=\cdots K_{t_{r}}\left(W_{1}, W_{2}\right)=K\left(W_{1}, W_{2}\right)
$$

The functions $K_{t_{1}}\left(w_{1}, w_{2}\right), K_{f}\left(w_{1}, w_{2}\right)$ depend on the fractional part $\{t\}$ and thus are periodic in time with period 1. Introduce the sums

$$
\begin{aligned}
Z_{t_{1}}\left(x ; a_{1}\right) & =\sum_{m \in \mathbb{Z}^{1}} K_{t_{1}}\left(x, a_{1}+m x_{0}\right) \\
Z_{t_{j}}\left(a_{j-1}, a_{j}\right) & =Z_{t_{j}}\left(a_{j-1}, a_{j}\right) \\
& =\sum_{m \in \mathbb{Z}^{1}} K_{t_{j}}\left(a_{j-1}, a_{j}+m x_{0}\right), \quad 2 \leqslant j \leqslant r \\
Z_{f}\left(a_{r}, a_{r+1}\right) & =\sum_{m \in \mathbb{Z}^{1}} K_{f}\left(a_{r}, a_{r+1}+m x_{0}\right)
\end{aligned}
$$

and the probabilities

$$
\begin{aligned}
& p_{t_{1}}\left(x, a_{1}+m_{1} x_{0}\right) \\
& \quad=Z_{t_{1}}^{-1}\left(x, a_{1}\right) K_{t_{1}}\left(x, a_{1}+m_{1} x_{0}\right) \\
& p_{t_{j}}\left(a_{j-1}+m_{j-1} x_{0}, a_{j}+m_{j} x_{0}\right) \\
& \quad=Z_{t_{j}}^{-1}\left(a_{j-1}, a_{j}\right) K_{t_{j}}\left(a_{j-1}+m_{j-1} x_{0}, a_{j}+m_{j} x_{0}\right), \quad 2 \leqslant j \leqslant r \\
& p_{f}\left(a_{r}+m_{r} x_{0} ; a_{r+1}+m_{r+1} x_{0}\right) \\
& \quad=Z_{f}^{-1}\left(a_{r}, a_{r+1}\right) K_{f}\left(a_{r}+m_{r} x_{0} ; a_{r+1}+m_{r+1} x_{0}\right)
\end{aligned}
$$

Consider the sequence of independent random variables $\xi_{1}$, $\xi_{2}, \ldots, \xi_{r+1}$, where each variable $\xi_{j}$ takes values $m \in \mathbb{Z}^{1}$, and $\xi_{1}$ has the distribution with the probabilities $p_{1}\left(x, a_{1}+m x_{0}\right)$, while $\xi_{j}, 2 \leqslant j \leqslant r$, have the distribution with the probabilities $p_{t_{j}}\left(a_{j-1}, a_{j}+m x_{0}\right)=$ $p\left(a_{j-1}, a_{j}+m x_{0}\right)$, which depends only on $\{t\}$ but not on $j$, and $\xi_{r+1}$ has the distribution with the probabilities $p_{f}\left(a_{r}, a_{r+1}+m x_{0}\right)$, which also depend only on $\{t\}$. Then the sum in (4) can be rewritten as follows:

$$
\begin{align*}
\sum= & \sum\left(a_{1}, a_{2}, \ldots, a_{r+1}\right) \\
= & \sum_{m_{1}, m_{2}, \ldots, m_{r+1}} K_{t_{1}}\left(x, a_{1}+m_{1} x_{0}\right) \\
& \times \prod_{j=2}^{r} K_{t_{1}}\left(a_{j-1}+m_{j-1} x_{0}, a_{j}+m_{j} x_{0}\right) \\
& \times K_{f}\left(a_{r}+m_{r} x_{0}, a_{r+1}+m_{r+1} x_{0}\right) e^{-v\left(a_{r+1}+m_{r+1} x_{0}\right)} \\
= & Z_{t_{1}}\left(x ; a_{1}\right) Z\left(a_{1}, a_{2}\right) \cdots Z\left(a_{r-1}, a_{r}\right) Z_{f}\left(a_{r}, a_{r+1}\right) \\
& \times \sum_{m_{1}, m_{2}, \ldots, m_{r+1}} p_{t_{1}}\left(x, a_{1}+m_{1} x_{0}\right) \\
& \times \prod_{j=2}^{r} p_{j}\left(a_{j-1}+m_{j-1} x_{0}, a_{j}+m_{j} x_{0}\right) \\
& \times p_{f}\left(a_{r}+m_{r} x_{0}, a_{r+1}+m_{r+1} x_{0}\right) e^{-v\left(a_{r+1}+m_{r+1} x_{0}\right)} \\
= & Z_{t_{1}}\left(x ; a_{1}\right) \prod_{j=2}^{r} Z_{j}\left(a_{j-1}, a_{j}\right) Z_{f}\left(a_{r}, a_{r+1}\right) \\
& \times \sum_{n_{1}, n_{r+1}} p_{t_{1}}\left(x, a_{1}+n_{1} x_{0}\right) \prod_{j=2}^{r} p_{j}\left(a_{j-1}, a_{j}+n_{j} x_{0}\right) \\
& \times p_{f}\left(a_{r}, a_{r+1}+n_{r+1} x_{0}\right) e^{-v\left(a_{r+1}+\left(n_{1}+\cdots+n_{r+1}\right) x_{0}\right)} \\
= & Z_{t_{1}}\left(x ; a_{1}\right) \prod_{j=2}^{r} Z\left(a_{j-1}, a_{j}\right) Z_{f}\left(a_{r}, a_{r+1}\right) \\
& \times E_{\xi} e^{-v\left(a_{r+1}+\left(\xi_{1}+\cdots+\xi_{r+1}\right) x_{0}\right)} \tag{5}
\end{align*}
$$

where $E_{\xi}$ is the expectation with respect to the joint distribution of the random variables $\xi_{j}, 1 \leqslant j \leqslant r+1$. Put $\mu\left(a_{1}\right)=E \xi_{1}, \mu\left(a_{j-1}, a_{j}\right)=E \xi_{j}$ for $\quad 1 \leqslant j \leqslant r+1, \quad d\left(a_{1}\right)=D\left(\xi_{1}\right)=E_{\xi}\left(\xi_{1}-\mu\left(a_{1}\right)\right)^{2}, \quad d\left(a_{j-1}, a_{j}\right)=E_{\xi}\left(\xi_{j}-\right.$ $\left.\mu\left(a_{j-1}, a_{j}\right)\right)^{2}, M=\mu\left(a_{1}\right)+\sum_{j=2}^{r+1} \mu\left(a_{j-1}, a_{j}\right)$, and $D=d\left(a_{1}\right)+\sum_{j=2}^{r+1} d\left(a_{j-1}, a_{j}\right)$.

Lemma 1. Under the conditions of Theorem 1, the sequence of
random variables $\xi_{1}, \xi_{2}, \ldots, \xi_{r+1}$ satisfies the central local limit theorem of probability theory in the form

$$
\begin{equation*}
\sup _{a_{1}, \ldots, a_{r+1}}\left|P_{\xi}\left(\xi_{1}+\cdots+\xi_{r+1}=m\right)-\frac{1}{(2 \pi D)^{1 / 2}} e^{-(m-M)^{2} / 2 D}\right| \leqslant \varepsilon_{t} \tag{6}
\end{equation*}
$$

where $\varepsilon_{t}$ tends to zero as $t \rightarrow \infty$.
The statement of Lemma 1 means that the convergence to zero of the difference in (6) is uniform in $m_{1}$ and $a_{1}, a_{2}, \ldots, a_{r+1}$. Lemma 1 can be easily proven by standard methods of probability theory (see, e.g., ref. 4). We omit the proof. During the proof one must keep in mind the boundedness of $|F|$ and $|B|$.

Consider in more detail the expectation

$$
E=E_{\xi} e^{-v_{0}\left(a_{r+1}+\left(\epsilon_{1}+\cdots+\xi_{r+1}\right) x_{0}\right)}
$$

In view of Lemma 1 and property 1 of $P_{0}$, it is equal to

$$
\begin{aligned}
E & =\sum_{m} e^{-v_{0}\left(a_{r}+1+m x_{0}\right)} P_{\xi}\left\{\xi_{1}+\cdots+\xi_{r+1}=m\right\} \\
& =\sum_{m} e^{-v_{0}\left(a_{r+1}+m x_{0}\right)} \frac{1}{(2 \pi D)^{1 / 2}} e^{-(m-M)^{2} / 2 D}+e^{c_{0}} \delta_{t}
\end{aligned}
$$

where $\delta_{t} \rightarrow 0$ as $t \rightarrow \infty$. Using property 3 of $P_{0}$, we easily get

$$
E=A+\delta_{t}^{(1)}
$$

where $\delta_{t}^{(1)} \rightarrow 0$ as $t \rightarrow \infty$ uniformly in $a_{1}, \ldots, a_{r+1}$. Thus,

$$
\begin{align*}
& \varphi(x ; t) \sim A \int \cdots \int d a_{1} d a_{2} \cdots d a_{r} d a_{r+1} Z_{t_{1}}\left(x ; a_{1}\right) \\
& \times \prod_{j=2}^{r} Z\left(a_{j-1}, a_{j}\right) Z_{f}\left(a_{r}, a_{r+1}\right) \tag{7}
\end{align*}
$$

The expression (7) can be studied with the methods of statistical mechanics. Consider $Z\left(a^{\prime}, a^{\prime \prime}\right)$ as a transfer matrix of a one-dimensional system and find its positive eigenvector $e(a)$ and the corresponding positive eigenvalue $\lambda$ :

$$
\int e\left(a^{\prime}\right) Z\left(a^{\prime}, a^{\prime \prime}\right) d a^{\prime}=\lambda e\left(a^{\prime \prime}\right)
$$

Introduce the Markov transition operator $\pi$ with the transition probabilities

$$
\pi\left(a^{\prime}, a^{\prime \prime}\right)=\frac{Z\left(a^{\prime}, a^{\prime \prime}\right) e\left(a^{\prime}\right)}{\lambda e\left(a^{\prime \prime}\right)}
$$

giving the density of the transition $a^{\prime \prime} \rightarrow a^{\prime}$. Then (7) can be rewritten as

$$
\begin{aligned}
\varphi(x ; t) \sim & A \lambda^{r} \int \cdots \int d a_{1} \cdots d a_{r} d a_{r+1} Z_{t_{1}}\left(x ; a_{1}\right)\left[e\left(a_{1}\right)\right]^{-1} \pi\left(a_{2}, a_{3}\right) \\
& \times \cdots \pi\left(a_{r-1}, a_{r}\right) Z\left(a_{r}, a_{r+1}\right) e\left(a_{r}\right) \\
= & A \lambda^{r} \int Z_{t_{1}}\left(x ; a_{1}\right)\left[e\left(a_{1}\right)\right]^{-1} \pi^{(r)}\left(a_{1}, a_{r}\right) \\
& \times Z\left(a_{r}, a_{r+1}\right) e\left(a_{r}\right) d a_{1} d a_{r} d a_{r+1}
\end{aligned}
$$

The ergodic theorem for the Markov chain generated by the operator $\pi$ shows that $\pi^{(r)}\left(a_{1}, a_{r}\right)$ asymptotically does not depend on $a_{r}$ and is exponentially close to the stationary distribution of this chain. Denote this distribution by $\pi\left(a_{1}\right)$. It is well known that it has the form $e\left(a_{1}\right) e^{*}\left(a_{1}\right)$, where $e^{*}\left(a_{1}\right)$ is the positive eigenvector of the adjoint operator, i.e.,

$$
\int Z\left(a^{\prime}, a^{\prime \prime}\right) e^{*}\left(a^{\prime \prime}\right) d a^{\prime \prime}=\hat{\lambda} e^{*}\left(a^{\prime}\right)
$$

Thus

$$
\varphi(x, t) \sim A \cdot A_{1} \lambda^{r} \int Z_{t_{1}}\left(x ; a_{1}\right) e^{*}\left(a_{1}\right) d a_{1}
$$

where $A_{1}=\iint Z\left(a_{r}, a_{r+1}\right) e\left(a_{r}\right) d a_{r} d a_{r+1}$.
Taking the derivative of the rhs of (4) with respect to $x$ and making the same analysis, we find a similar expression for $\varphi_{x}$ :

$$
\varphi_{x}(x, t) \sim A A_{1} \lambda^{r} \int \frac{\partial}{\partial x} Z_{t_{1}}\left(x ; a_{1}\right) e\left(a_{1}\right) d a_{1}
$$

Finally we get that for $t \rightarrow \infty$

$$
\begin{equation*}
u=-2 \mu \frac{\varphi_{x}(x ; t)}{\varphi(x ; t)} \sim 2 \mu \frac{\int\left[\partial Z_{t_{1}}\left(x ; a_{1}\right) / \partial x\right] e\left(a_{1}\right) d a_{1}}{\int Z_{t_{1}}\left(x ; a_{1}\right) e\left(a_{1}\right) d a_{1}} \tag{8}
\end{equation*}
$$

The rhs of (8) is a solution of (1) periodic in space and time and (8) gives the assertion of Theorem 1.

It is clear that the properties of smoothness of

$$
u^{(0)}(x ; t)=-2 \mu \frac{\int\left[\partial Z_{t_{1}}\left(x ; a_{1}\right) / \partial x\right] e\left(a_{1}\right) d a_{1}}{\int Z_{t_{1}}\left(x ; a_{1}\right) e\left(a_{1}\right) d a_{1}}
$$

depend on the smoothness of $F$. In particular, if $F(x) \in C^{k}\left(S^{1}\right)$, then $u^{(0)}(x ; t) \in C^{k-1}\left(S^{1}\right)$ for any $t$ and one can prove easily the convergence of $\left(\partial^{i} / \partial x^{i}\right) u(x ; t), \quad i \leqslant k-1$, to $\left(\partial^{i} / \partial x^{i}\right) u^{(0)}(x ; t)$. Also, the convergence in Theorem 1 is uniform in $x$ on any compact subset of $R^{1}$. Certainly in general it cannot be uniform on the whole line, because of fluctuations of $v$.

## 3. PROOF OF THEOREM 2

Again we use the Cole-Hopf substitution, which now gives the expres$\operatorname{sion}$ for $\varphi$ in the form

$$
\begin{align*}
\varphi(x ; t)= & \int_{-\infty}^{\infty} d y\{\exp [-v(y)]\} \\
& \times\left\{\exp \left[\int_{0}^{t} F(w(\tau)) d b(\tau)\right]\right\} d \Pi_{(x, y)}^{(t, 0)}(W) \tag{9}
\end{align*}
$$

Here $\int_{0}^{\tau} F(w(\tau)) d b(\tau)$ is a stochastic integral, and $B(\tau)=d b(\tau) / d \tau$ is white noise. It is worthwhile to stress that $\{w(\tau)\}$ and $\{b(\tau, 0)\}$ are statistically independent Brownian motions. Therefore $\varphi(x ; t)$ is random because of the randomness of $b$. We proceed in the same way as before. Take an integer $r=r(t)$ for which $r / t \rightarrow 1$ as $t \rightarrow \infty$ and divide the interval ( $0, t$ ) into $r$ equal parts. Denote the points of the division by $t=t_{0}>t_{1}>\cdots>t_{r}=0_{r}$ and rewrite (9) as follows:

$$
\begin{align*}
\varphi(x ; t)= & \int \cdots \int d a_{1} \cdots d a_{r} \sum_{\substack{m_{1}, \ldots, m_{r} \\
m_{j} \in \mathbb{Z}^{1}}} K_{1}\left(x, a_{1}+m_{1} x_{0}\right) \\
& \times K_{j}\left(a_{j-1}+m_{j-1} x_{0}, a_{j}+m_{j} x_{0}\right) \exp \left[-v\left(a_{r}+m_{r} x_{0}\right)\right] \tag{10}
\end{align*}
$$

where

$$
K_{j}\left(a^{\prime}, a^{\prime \prime}\right)=\int\left\{\exp \left[\int_{t_{j}}^{t_{j-1}} F(w(\tau)) d b(\tau)\right]\right\} d \Pi_{a^{\prime}, a^{\prime \prime}}^{\left(t_{j}, t_{j}\right)}(W)
$$

In the case of white noise the operators $K_{j}\left(a^{\prime}, a^{\prime \prime}\right)$ are random and statistically independent in a natural sense for different $j$. The periodicity of $F$ in $x$ implies

$$
K_{j}\left(a^{\prime}, a^{\prime \prime}\right)=K_{j}\left(a^{\prime}+m x_{0}, a^{\prime \prime}+m x_{0}\right), \quad m \in \mathbb{Z}^{1}
$$

This gives again a possibility to reduce the summation $\sum_{m_{1}, \ldots, m_{r}}$ to a problem concerning independent differently distributed random variables. Namely, introduce the partition functions

$$
\begin{aligned}
Z_{1}\left(x, a_{1}\right) & =\sum_{m} K_{1}\left(x, a_{1}+m x_{0}\right) \\
Z_{j}\left(a_{j-1}, a_{j}\right) & =\sum_{m \in \mathbb{Z}^{1}} K_{j}\left(a_{j-1}, a_{j}+m x_{0}\right)
\end{aligned}
$$

and the corresponding probabilities

$$
\begin{aligned}
& p_{1}\left(x, a_{1}+m_{1} x_{0}\right)=Z_{1}^{-1}\left(x, a_{1}\right) K\left(x, a_{1}+m_{1} x_{0}\right) \\
& p_{j}\left(a_{j-1}+m_{j-1} x_{0}, a_{j}+m_{j} x_{0}\right) \\
& \quad=Z_{j}^{-1}\left(a_{j-1}, a_{j}\right) K_{j}\left(a_{j-1}+m_{j-1} x_{0}, a_{j}+m_{j} x_{0}\right) \\
& \quad=Z_{j}^{-1}\left(a_{j-1}, a_{j}\right) K_{j}\left(a_{j-1}, a_{j}+\left(m_{j}-m_{j-1}\right) x_{0}\right)
\end{aligned}
$$

Then (10) takes the form

$$
\begin{aligned}
\varphi(x ; t)= & \int d a_{1} d a_{2} \cdots d a_{r} Z_{1}\left(x, a_{1}\right) Z_{2}\left(a_{1}, a_{2}\right) \cdots Z_{r}\left(a_{r-1}, a_{r}\right) \\
& \times \sum_{n_{1}, n_{2}, \ldots, n_{r} \in \mathbb{Z}^{1}} p_{1}\left(x, a_{1}+n_{1} x_{0}\right) \\
& \times p_{2}\left(a_{1}, a_{2}+n_{2} x_{0}\right) \cdots p_{r}\left(a_{r-1}, a_{r}+n_{r} x_{0}\right) \\
& \times \exp \left\{-r\left[a_{r}+\left(n_{1}+\cdots+n_{r}\right) x_{0}\right]\right\}
\end{aligned}
$$

Let $\xi_{1}, \ldots, \xi_{r}$ be $r$ independent integer-valued random variables where each $\xi_{j}$ has the distribution $p_{j}\left(a_{j-1}, a_{j}+m x_{0}\right), a_{0}=x$. We can write

$$
\begin{align*}
& \quad \sum_{n_{1}, n_{2}, \ldots, n_{r} \in \mathbb{Z}^{1}} p_{1}\left(x, a_{1}+n_{1} x_{0}\right) p_{2}\left(a_{1}, a_{2}+n_{2} x_{0}\right) \cdots p_{r}\left(a_{r-1}, a_{r}+n_{r} x_{0}\right) \\
& \quad \times \exp \left\{-r\left[a_{r}+\left(n_{1}+\cdots+n_{r}\right) x_{0}\right]\right\} \\
& \quad=E_{\xi} \exp \left\{-r\left[a_{r}+\left(\xi_{1}+\cdots+\xi_{r}\right) x_{0}\right]\right\} \tag{11}
\end{align*}
$$

Again as in Section 2 we encounter two problems. The first one concerns the validity of the local central limit theorem of probability theory, while the second one consists of the possibility of replacing the average (11) by its mathematical expectation $A$. Since the distribution $P_{0}$ has the properties $1-3$ (see Section 1), the second problem is simple because the local central limit theorem and the stability of the averages (see property 3 ) of the distribution $P_{0}$ show that (11) is equivalent to $A$ as $t \rightarrow \infty$.

In order to study the local central limit theorem, introduce

$$
\begin{gathered}
\mu_{1}\left(a_{1}\right)=E \xi_{1}, \quad \mu_{j}\left(a_{j-1}, a_{j}\right)=E \xi_{j}, \quad d_{1}\left(a_{1}\right)=D\left(\xi_{1}\right) \\
d_{j}\left(a_{j-1}, a_{j}\right)=D\left(\xi_{j}\right), \quad 2 \leqslant j \leqslant r, \quad \mathscr{M}_{r}=\mu_{1}\left(a_{1}\right)+\sum_{j=2}^{r} \mu_{j}\left(a_{j-1}, a_{j}\right) \\
D_{r}=d(a)+\sum_{j=2}^{r} d_{j}\left(a_{j-1}, a_{j}\right)
\end{gathered}
$$

Certainly, $\mathscr{M}_{r}$ and $D_{r}$ are andom variables, since they are functions of $b$.
Let $t \rightarrow \infty$. Consider the probability $P_{b}(t)$ (with respect to $b$ ) that the random variables $\xi_{1}, \xi_{2}, \ldots, \xi_{r}$ satisfy the local central limit theorem (lclt) in the form described in the Lemma 1.

Lemma 3. $\quad P_{b}(t) \rightarrow 1$ as $t \rightarrow \infty$.
The proof of the lemma is simple and we shall describe only the main steps. It uses characteristic functions. It is easy to show that there exists a finite covering of $S^{1}$ by arcs $C_{0}, C_{1}, C_{2}, \ldots, C_{s}, p>0, \delta>0$, such that $C_{0}$ is a symmetric neighborhood of 1 and for any $C_{j}, 1 \leqslant j \leqslant s$, the probability (with respect to $B$ ) that the characteristic function has on $C_{j}$ the absolute value less than $1-\delta$ is greater than $p$. This gives easily an exponential estimation for the characteristic function of the sum $\sum_{j-1}^{r} \xi_{j}$ outside a small neighborhood of 1 . The rest follows the traditional way of proving the local central limit theorem. ${ }^{(4)}$

Thus, under the conditions of Theorem 1 and for those $b$ for which the lclt is true we can write again

$$
\begin{equation*}
\varphi(x ; t) \sim A \int d a_{1} d a_{2} \cdots d a_{r} Z_{1}\left(x, a_{1}\right) \prod_{j=2}^{r} Z_{j}\left(a_{j-1}, a_{j}\right) \tag{12}
\end{equation*}
$$

Now $Z_{j}\left(a_{j-1}, a_{j}\right)$ are $b$-independent random variables. The analysis of (12) can be done again with the help of the theory of non-homogeneous Markov chains.

Namely, consider the conditional probabilities

$$
\begin{aligned}
\pi_{1}\left(a_{1} / a_{2}\right) & =\frac{Z_{2}\left(a_{1}, a_{2}\right)}{\int Z_{2}\left(a_{1}, a_{2}\right) d a_{1}} \\
\pi_{j}\left(a_{j} / a_{j+1}\right) & =\frac{\int Z_{2}\left(a_{1}, a_{2}\right) \cdots Z_{j}\left(a_{j-1}, a_{j}\right) Z_{j+1}\left(a_{j}, a_{j+1}\right) d a_{1} \cdots d a_{j-1}}{\int Z_{2}\left(a_{1}, a_{2}\right) \cdots Z_{j}\left(a_{j-1}, a_{j}\right) Z_{j+1}\left(a_{j}, a_{j+1}\right) d a_{1} \cdots d a_{j}}
\end{aligned}
$$

We can use them to rewrite the rhs of (12) in another way:

$$
\begin{gathered}
\varphi(x ; t) \sim A \Xi_{r} \int Z_{1}\left(x, a_{1}\right) \pi_{1}\left(a_{1} \mid a_{2}\right) \pi_{2}\left(a_{2} \mid a_{3}\right) \cdots \\
\times \pi_{r-1}\left(a_{r-1} \mid a_{r}\right) \pi_{r}\left(a_{r}\right) d a_{1} \cdots d a_{r}
\end{gathered}
$$

where

$$
\Xi_{r}=\int d a_{1} \cdots d a_{r} Z_{2}\left(a_{1}, a_{2}\right) \cdots Z_{r}\left(\dot{a}_{r-1}, a_{r}\right)
$$

plays the role of partition function. For the derivative $\varphi_{x}(x ; t)$ we have a similar expression:

$$
\begin{aligned}
& \varphi_{x}(x ; t) \sim A \Xi_{r} \int \frac{\partial Z_{1}}{\partial x}\left(x, a_{1}\right) \pi_{1}\left(a_{1} \mid a_{2}\right) \pi_{2}\left(a_{2} \mid a_{3}\right) \cdots \\
& \times \pi_{r-1}\left(a_{r-1} \mid a_{r}\right) \pi\left(a_{r}\right) d a_{1} \cdots d a_{r}
\end{aligned}
$$

Therefore this yields for the solution $u(x ; t)$ of the Burgers equation

$$
\begin{aligned}
u(x, t)= & -2 \mu \frac{\varphi_{x}(x ; t)}{\varphi(x ; t)} \sim \frac{-2 \mu \int\left(\partial Z_{1} / \partial x\right)\left(x, a_{1}\right) \pi_{1}\left(a_{1} \mid a_{2}\right) \cdots}{\int Z_{1}\left(x, a_{1}\right) \pi_{1}\left(a_{1} \mid a_{2}\right) \cdots} \\
& \times \frac{\cdots \pi_{r-1}\left(a_{r-1} \mid a_{r}\right) \pi\left(a_{r}\right) d a_{1} \cdots d a_{r}}{\cdots \pi_{r-1}\left(a_{r-1} \mid a_{r}\right) \pi\left(a_{r}\right) d a_{1} \cdots d a_{r}} \\
= & \frac{\int\left(\partial Z_{1} / \partial x\right)\left(x, a_{1}\right) \pi_{1}\left(a_{1} \mid a_{k}\right) \pi_{k}\left(a_{k+1} \mid a_{k}\right) \cdots}{\int Z_{1}\left(x, a_{1}\right) \pi_{1}\left(a_{1} \mid a_{k}\right) \pi_{k}\left(a_{k+1} \mid a_{k}\right) \cdots} \\
& \times \frac{\pi_{r-1}\left(a_{r-1} \mid a_{r}\right) \pi\left(a_{r}\right) d a_{1} d a_{k} \cdots d a_{r}}{\pi_{r-1}\left(a_{r-1} \mid a_{r}\right) \pi\left(a_{r}\right) d a_{1} d a_{k} \cdots d a_{r}}
\end{aligned}
$$

for any $k$. Here $\pi\left(a_{1} \mid a_{k}\right)$ is the conditional density corresponding to the joint probability density

$$
\frac{Z_{2}\left(a_{1}, a_{2}\right) \cdots Z_{r}\left(a_{r-1}, a_{r}\right) d a_{1} d a_{2} \cdots d a_{r}}{\Xi}
$$

Now we remark that for large $k$ the conditional distribution $\pi\left(a_{1} \mid a_{k}\right)$ becomes almost independent of $a_{k}, a_{k+1}, \ldots, a_{r}$ and thus independent of $B(\tau), 0 \leqslant \tau \leqslant t_{k}$. This follows easily from the ergodic theorem for Markov chains. To be more precise, let us formulate the following lemma.

Lemma 4. There exist positive constants $\rho<1$ and $C_{2}<\infty$ and events $\quad S_{k} \in B(0, k), \quad k=1,2, \ldots, \quad P_{b}\left(S_{k}\right)>1-C_{2} \rho^{k} \quad$ and $\quad$ a functional
$H_{x}^{(k)}\left(b\left(t_{1}, t_{2}\right), 0 \leqslant t_{1}, t_{2} \leqslant k\right)$ defined on $B(0, k)$ such that if $\left(b\left(t_{1}, t_{2}\right)\right.$, $\left.t-k \leqslant t_{1}<t_{2} \leqslant t\right) \in S_{k}$, then

$$
\left|u(x, t)-H_{x}^{(k)}\left(b\left(t_{1}, t_{2}\right), t-k \leqslant t_{1}<t_{2} \leqslant t\right)\right| \leqslant C_{2} \rho^{k}
$$

The functional $H_{x}^{(k)}$ is a periodic function of $x$ of period $x_{0}$.
The proof of the lemma goes as follows. The transition densities $\pi_{j}\left(a_{j-1} \mid a_{j}\right)$ are bigger than some constant $\sigma>0$ with a positive probability. It is easy to show that with the probability not less than $1-C_{2} \rho^{k}$ the number of such $j$ is bigger then $\beta k$ for some $\beta>0$. Then the conditional distribution $\pi\left(a_{1} \mid a_{k}\right)$ does not depend on $k$. The periodicity of $H_{x}^{(k)}$ on $x$ follows easily from the expressions for $\varphi(x ; t)$.

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