

# FINITELY FIXED IMPLIES LOOSELY BERNOULLI, A DIRECT PROOF

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## ABSTRACT

As defined in the literature, a process is loosely Bernoulli if a certain property  $P(\varepsilon)$  is satisfied for every  $\varepsilon > 0$ . Using only facts about stationary joinings of processes, it is shown that given  $\varepsilon > 0$  there exists  $\delta > 0$  such that whenever two processes are separated by less than  $\delta$  in the  $\bar{f}$ -metric and one of them is loosely Bernoulli, the other is "almost" loosely Bernoulli in the sense that  $P(\varepsilon)$  is satisfied. As easy corollaries, one has that loosely Bernoulli processes are closed in the  $\bar{f}$ -metric and that finitely fixed processes are loosely Bernoulli.

## 1. Introduction

It was shown in [3] that one can give a very short proof of the fact that Ornstein's condition "finitely determined" implies "very weak Bernoulli" based on the following theorem.

**THEOREM [3; corollary 1].** *Let  $B$  be a finite set, and let  $U = \{U_i\}_{i=-\infty}^{\infty}$  and  $V = \{V_i\}_{i=-\infty}^{\infty}$  be stationary stochastic processes with values in the space  $B^{\mathbb{Z}}$  of doubly infinite sequences from  $B$ . Suppose, moreover, that  $U$  is very weak Bernoulli. Then if the processes  $U$  and  $V$  are within  $\varepsilon$  in the  $\bar{d}$ -metric, there exists a positive integer  $m$  such that*

$$E\bar{d}_m(\text{dist } V_1^m, \text{dist}(V_1^m | V_{-\infty}^0)) < 2\varepsilon$$

where  $E$  indicates expectation,  $\text{dist } V_1^m$  denotes the distribution of  $V_1^m = (V_1, \dots, V_m)$  and  $\text{dist}(V_1^m | V_{-\infty}^0)$  denotes the conditional distribution of  $V_1^m$  given the past  $V_{-\infty}^0 = (\dots, V_{-1}, V_0)$ , considered as a function of  $V_{-\infty}^0$ .

In this article, we derive the corresponding result for the  $\bar{f}$ -metric and apply it to show that finitely fixed implies loosely Bernoulli. First, for the benefit of those

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readers who are familiar with the relevant  $\bar{d}$ -concepts but who may not be familiar with [3], we provide a thumbnail sketch of the proof of the above theorem as a reference point in following the rather more involved  $\bar{f}$ -arguments which follow. The necessary  $\bar{f}$ -definitions and notation are presented in the next section, followed by the proof that  $FF \Rightarrow LB$  in §3, and finally in §4 the proof of the main result: the  $\bar{f}$ -version of the above theorem.

SKETCH OF PROOF. The result depends on two facts. First, we can assume  $U$  and  $V$  arise from a jointly stationary process  $(U, V)$  such that  $Ed_1(U_0, V_0) = \bar{d}(\text{dist } U, \text{dist } V)$ . Second, by a result of Rohlin [6, p. 66] in the jointly stationary process, no additional information is gained about  $U_0$  from the remote past  $V_{-\infty}^-$ , provided we know the total past  $U_{-\infty}^-: \lim_{j \rightarrow \infty} h(U_0 | U_{-\infty}^- V_{-\infty}^-) = h(U_0 | U_{-\infty}^-)$ , where  $h$  represents entropy. From this it follows that

$$\bar{d}_1(\text{dist}(U_0 | U_{-\infty}^- V_{-\infty}^-), \text{dist}(U_0 | U_{-\infty}^-)) \rightarrow 0 \quad \text{as } j \rightarrow \infty.$$

Since  $U$  is VWB, we choose  $k$  so that  $E\bar{d}_k(\text{dist } U_1^k, \text{dist}(U_1^k | U_{-\infty}^0)) < \varepsilon$ . Let  $\pi$  represent the  $j$ -fold product of the distribution  $U_1^k$  with itself. By using the sub-additivity of the  $\bar{d}$ -metric over the disjoint blocks of length  $k$  and then applying stationarity [3], we conclude that

$$\begin{aligned} E\bar{d}_{jk}(\text{dist } U_1^k, \pi) &\leq j^{-1} \sum_{i=1}^{j-1} E\bar{d}_k(\text{dist}(U_{ik+1}^{ik+k} | U_1^k), \text{dist } U_1^k) \\ &= j^{-1} \sum_{i=1}^{j-1} E\bar{d}_k(\text{dist}(U_1^k | U_{1-ik}^0), \text{dist } U_1^k) \end{aligned}$$

which approaches  $E\bar{d}_k(\text{dist}(U_1^k | U_{-\infty}^0), \text{dist } U_1^k)$  as  $j \rightarrow \infty$ .

Similarly,

$$E\bar{d}_{jk}(\text{dist}(U_1^k | U_{-\infty}^0 V_{-\infty}^0), \pi) \leq j^{-1} \sum_{i=0}^{j-1} E\bar{d}_k(\text{dist}(U_1^k | U_{-\infty}^0 V_{-\infty}^{-ik}), \text{dist } U_1^k),$$

which, by our earlier remarks, approaches  $E\bar{d}_k(\text{dist}(U_1^k | U_{-\infty}^0), \text{dist } U_1^k)$  as  $j \rightarrow \infty$ . Hence by the triangle inequality, for  $j$  sufficiently large,

$$(0) \quad E\bar{d}_{jk}(\text{dist } U_1^k, \text{dist}(U_1^k | U_{-\infty}^0 V_{-\infty}^0)) < 2\varepsilon.$$

On the other hand, by definition of the  $\bar{d}$ -distance we conclude that

$$E\bar{d}_{jk}(\text{dist } U_1^k, \text{dist } V_1^k) \leq Ed_{jk}(U_1^k, V_1^k) = Ed_1(U_0, V_0) < \varepsilon.$$

Similarly, we get that

$$E\bar{d}_{jk}(\text{dist}(U_1^k | U_{-\infty}^0 V_{-\infty}^0), \text{dist}(V_1^k | U_{-\infty}^0 V_{-\infty}^0)) < \varepsilon.$$

By the triangle inequality, in (0) we can replace  $U_1^k$  by  $V_1^k$ , provided we replace “ $2\varepsilon$ ” by “ $4\varepsilon$ ”. Dropping the condition on  $U_{-\infty}^0$ , we get our result.

For the  $\bar{f}$ -case, the approach is similar but more involved, since in the stationary joining which realizes the  $\bar{f}$ -distance between  $U$  and  $V$ , the processes  $U$  and  $V$  are not themselves jointly stationary. Instead,  $U$  and  $V$  must be recovered from the jointly stationary processes  $\bar{U}, \bar{V}$  by the non-stationary procedure of arbitrarily assigning a time-zero and concatenating  $\bar{U}, \bar{V}$  outputs. Conversely, the transition from  $U, V$  to  $\bar{U}, \bar{V}$  must be effected by means of marker processes  $M_U, M_V$  which delineate the beginnings of blocks of  $U, V$  outputs which are paired in the  $\bar{f}$ -match.

**2. Preliminaries**

For  $S$  countable, let  $S^\infty$  denote the set of all infinite sequences  $x = (x_i)_{i=-\infty}^\infty$  from  $S$ . We make  $S^\infty$  a measurable space by adjoining the usual product  $\sigma$ -algebra generated by the partition of  $S$  into discrete points. By a process we mean a measurable map  $X$  from some measurable space  $\Omega$  to  $S^\infty$ . If  $X : \Omega \rightarrow S^\infty$  is a process and  $i$  is an integer,  $X_i$  denotes the map from  $\Omega$  to  $S$  such that  $X_i(\omega) = X(\omega)_i, \omega \in \Omega$ . For integers  $m, n$  with  $m \leq n$ ,  $X_m^n$  denotes the function  $(X_m, X_{m+1}, \dots, X_n)$ ;  $X_{-\infty}^n$ , the function  $(\dots, X_{n-1}, X_n)$ ; and if  $n > 0$ ,  $X^n$  denotes  $(X_1, \dots, X_n)$ . If  $(\Omega, \mathcal{F})$  is a measurable space, let  $\mathcal{P}(\Omega)$  be the family of all probability measures on  $\mathcal{F}$ . If  $X_1, \dots, X_n$  are measurable maps from  $\Omega$  to measurable spaces  $S_1, \dots, S_n$ , respectively, and if  $P \in \mathcal{P}(\Omega)$ , then  $P(\cdot | X_1, \dots, X_n)$  denotes a map from  $\Omega$  to  $\mathcal{P}(\Omega)$  such that for each set  $E \in \mathcal{F}$ , the random variable  $P(E | X_1, \dots, X_n)$  serves as a conditional expectation under  $P$  of the characteristic function of  $E$  given the sub- $\sigma$ -algebra of  $\mathcal{F}$  generated by  $X_1, \dots, X_n$ . If in addition  $X$  is a measurable map from  $\Omega$  to a measurable space  $(S, \mathcal{S})$ , then  $P^X(\cdot | X_1, \dots, X_n)$  denotes the  $\mathcal{P}(S)$ -valued map defined on  $\Omega$  such that for each  $E \in \mathcal{S}$  we have

$$P^X(E | X_1, \dots, X_n) = P(\{X \in E\} | X_1, \dots, X_n).$$

The symbol  $P^X$  denotes the distribution of  $X$ , i.e., the probability measure on  $\mathcal{S}$  such that  $P^X(E) = P(X \in E), E \in \mathcal{S}$ .

For a finite set  $R$ , let  $\hat{R} = \bigcup_{n=1}^\infty R^n$ , the collection of all finite sequences from  $R$ . Since  $R$  is countable, we may apply the comments of the paragraph above and consider the measurable space  $\hat{R}^\infty$ , processes  $\hat{X} : \Omega \rightarrow \hat{R}^\infty$ , etc.

For the rest of the paper, fix a finite set  $B$ . Let  $T_B$  denote the shift on  $B^\infty$ ;  $T_{\hat{B}}$ , the shift on  $\hat{B}^\infty$ . We shall deal with distributions on  $B^\infty$  and  $\hat{B}^\infty$  which are ergodic and have finite entropy.

We shall consider two versions of the  $\bar{f}$ -metric, which is described in detail in [1, 2, 7]. For  $b \in \hat{B}$ , let  $l(b)$  denote its length. Define  $f : \hat{B} \times \hat{B} \rightarrow [0, \infty)$  so that

$$f(b, c) = [l(b) + l(c) - 2l(\text{longest monotone match in } b, c)]/2.$$

It is easily shown that  $f$  is a metric. For  $n = 1, 2, \dots$ , define  $\hat{f}_n : \hat{B}^n \times \hat{B}^n \rightarrow [0, \infty)$  so that  $\hat{f}_n(b, c) = f(\hat{b}, \hat{c})/n$ , where  $\hat{b}, \hat{c}$  are the elements in  $\hat{B}$  obtained from  $b, c$  by concatenating. Let  $\bar{f}_n$  be the restriction of  $\hat{f}_n$  to  $B^n \times B^n$ . The sequence  $\{\hat{f}_n\}$  is subadditive in the following sense: let  $(x_1, \dots, x_{im}), (y_1, \dots, y_{im}) \in \hat{B}^{im} \times \hat{B}^{im}$ , then

$$\hat{f}_{im}((x_1, \dots, x_{im}), (y_1, \dots, y_{im})) \leq \sum_{j=0}^{i-1} \hat{f}_m((x_{jm+1}, \dots, x_{j+m}), (y_{jm+1}, \dots, y_{j+m})).$$

Similarly for the sequence  $\{\bar{f}_n\}$ . If  $A$  is countable and  $\rho : A^n \times A^n \rightarrow [0, \infty)$  is given, then if  $\mu, \nu$  are probability measures on  $A^n$ , the symbol  $\rho_n(\mu, \nu)$  denotes  $\inf_{(X,Y)} E\rho_n(X, Y)$ , where  $E$  indicates the expectation and the infimum is over all random variables  $X, Y$  which are  $A^n$ -valued with  $\text{dist } X = \mu, \text{dist } Y = \nu$ . If  $\nu, \mu$  are stationary probability measures on  $B^\infty$ ,  $\bar{f}(\mu, \nu)$ , the  $\bar{f}$ -distance between them, is defined to be  $\limsup_{n \rightarrow \infty} \bar{f}_n(\mu_n, \nu_n)$ , where  $\mu_n, \nu_n$  are the  $n$ th order marginal distributions of  $\mu, \nu$ , respectively.

Finally, if  $V$  is an ergodic process with state space  $B$  and distribution  $\nu$ , we say that  $\nu$  is loosely Bernoulli (LB) [1, 2] if for every  $\epsilon > 0$  there exists an integer  $m$  such that  $E_\nu \bar{f}(\nu^{\nu^m}, \nu^{\nu^m}(\cdot | V_{-\infty}^0)) < \epsilon$ .

### 3. Results

**THEOREM.** *Let  $Y, X$  be ergodic processes with state space  $B$  and distributions  $\mu, \nu$ . Let  $\mu$  be LB. Then if  $\bar{f}(\mu, \nu) < \epsilon^2$ , for all  $k$  sufficiently large we have  $E_\nu \bar{f}_k(\nu^{X^k}, \nu^{X^k}(\cdot | X_{-\infty}^0)) < 84\epsilon/(1 - \epsilon)$ .*

**COROLLARY.** *If  $\{\mu_i\}_{i=1}^\infty, \mu$  are  $T_B$ -ergodic and if each  $\mu_i, i = 1, 2, \dots$  is LB and if the  $\mu_i$  converge to  $\mu$  in the  $\bar{f}$ -metric, then  $\mu$  is LB.*

**DEFINITION.** We say that a  $T_B$ -ergodic measure  $\mu$  is finitely fixed (FF) [7] if convergence of any sequence  $\{\mu_i\}$  of  $T_B$ -ergodic measures to  $\mu$  both weakly and in entropy implies convergence of the  $\mu_i$  to  $\mu$  in the  $\bar{f}$ -metric.

**COROLLARY.** *If a  $T_B$ -ergodic measure  $\mu$  is FF, then  $\mu$  is LB.*

**PROOF.** The  $m$ th order Markov approximants of  $\mu$  converge to  $\mu$  weakly and in entropy and are LB. Hence  $\mu$  is LB.

REMARK. The conclusions of the two corollaries are already known. However, as Weiss points out [7, p. 0.2, p. 6.5], the proof of these facts is unsatisfactory from an aesthetic point of view, since it involves a detour through the equivalence theorem and the  $\bar{d}$  result of Ornstein and Weiss that finitely determined processes are very weak Bernoulli [5]. Our proof uses only properties of the stationary joining discussed in [7, prop. 2.6] and is a modification of a recent short proof of the Ornstein-Weiss result [3].

4. In this section we present the proof of the main theorem.

In the following if  $\mu, \nu$  are probability measures on a common space,  $|\mu - \nu|$  denotes the total variation distance between them.

LEMMA 1. *Let  $P, Q$  be probability measures on a measurable space  $(\Omega, \mathcal{F})$ , where  $\mathcal{F}$  is countably generated. Let  $\mathcal{G}$  be any sub- $\sigma$ -field of  $\mathcal{F}$ . Then*

$$E_P |P(\cdot | \mathcal{G}) - Q(\cdot | \mathcal{G})| \leq 4|P - Q|.$$

PROOF. Routine.

As a corollary we get

LEMMA 2. *Let  $P, Q, (\Omega, \mathcal{F})$ , and  $\mathcal{G}$  be as in Lemma 1. Let  $X_1, \dots, X_k$  be measurable functions from  $\Omega$  to  $B$ . Then*

$$|E_P \bar{f}_k(P^{X^k}, P^{X^k}(\cdot | \mathcal{G})) - E_Q \bar{f}_k(Q^{X^k}, Q^{X^k}(\cdot | \mathcal{G}))| \leq 6|P - Q|.$$

PROOF. Clearly

$$\begin{aligned} &|\bar{f}_k(P^{X^k}, P^{X^k}(\cdot | \mathcal{G})) - \bar{f}_k(Q^{X^k}, Q^{X^k}(\cdot | \mathcal{G}))| \\ &\leq \bar{f}_k(P^{X^k}, Q^{X^k}) + \bar{f}_k(P^{X^k}(\cdot | \mathcal{G}), Q^{X^k}(\cdot | \mathcal{G})). \end{aligned}$$

From Lemma 1 we get that

$$\begin{aligned} &|E_P \bar{f}_k(P^{X^k}, P^{X^k}(\cdot | \mathcal{G})) - E_Q \bar{f}_k(Q^{X^k}, Q^{X^k}(\cdot | \mathcal{G}))| \\ &\leq E_P |\bar{f}_k(P^{X^k}, P^{X^k}(\cdot | \mathcal{G})) - \bar{f}_k(Q^{X^k}, Q^{X^k}(\cdot | \mathcal{G}))| \\ &\quad + |E_P \bar{f}_k(Q^{X^k}, Q^{X^k}(\cdot | \mathcal{G})) - E_Q \bar{f}_k(Q^{X^k}, Q^{X^k}(\cdot | \mathcal{G}))| \\ &\leq |P - Q| + 4|P - Q| + |P - Q| = 6|P - Q|. \end{aligned}$$

Here we used the fact that the total variation distance upper bounds the  $\bar{f}$ -distance (since it upper bounds the  $\bar{d}$ -distance).

LEMMA 3. *Let  $A$  be a countable set. For each  $n = 1, 2, \dots$ , let a metric  $\rho_n : A^n \times A^n \rightarrow [0, \infty)$  be given. Suppose the sequence  $\{\rho_n\}$  is subadditive. Let  $X, Y$*

be jointly stationary processes defined on  $(\Omega, \mathcal{F}, P)$ , where  $X$  has state space  $A$ . Then for any  $m = 1, 2, \dots$  we have

$$(1) \quad \limsup_{i \rightarrow \infty} E\rho_{im}(P^{X^{im}}, P^{X^{im}}(\cdot | X_{-\infty}^0, Y_{-\infty}^0)) \leq 2E\rho_m(P^{X^m}, P^{X^m}(\cdot | X_{-\infty}^0)) \\ + \limsup_{i \rightarrow \infty} E\rho_m(P^{X^m}(\cdot | X_{-\infty}^0), P^{X^m}(\cdot | X_{-\infty}^0, Y_{-\infty}^{-i})).$$

Moreover, if  $A$  is finite and  $\rho_n(x, x) = 0$  for every  $x \in A^n$ , then the second term on the right-hand side of (1) is zero.

PROOF. See [3, proof of theorem 1].

The following result is easily proved using the fact that  $\bar{f}$  and  $\hat{f}$  are metrics.

LEMMA 4. Let  $\tilde{U}$  be a stationary process on  $(\Omega, \mathcal{F}, P)$  with state space  $\hat{B}$ . Let  $U$  be the nonstationary process with state space  $B$  obtained by "concatenating" the  $\tilde{U}$  output, i.e.,  $U_1^\infty$  is obtained by concatenating  $\tilde{U}_1, \tilde{U}_2, \dots$ , while  $U_{-\infty}^0$  comes from  $\dots, \tilde{U}_{-1}, \tilde{U}_0$ . Then for any  $n = 1, 2, \dots$  and any sub- $\sigma$ -fields  $\mathcal{G}_1, \mathcal{G}_2$  we have

$$|E\hat{f}_m(P^{\tilde{U}^m}(\cdot | \mathcal{G}_1), P^{\tilde{U}^m}(\cdot | \mathcal{G}_2)) - E\bar{f}_m(P^{U^m}(\cdot | \mathcal{G}_1), P^{U^m}(\cdot | \mathcal{G}_2))| \leq EI(\tilde{U}_0) - 1.$$

Finally, we come to the main result. Let  $\{X_i\}$  denote the projections from  $B^\infty$  to  $B$ .

THEOREM. Let  $\mu, \nu$  be ergodic on  $B^\infty$ . Let  $\bar{f}(\mu, \nu) < \varepsilon^2$ . Let  $\mu$  be loosely Bernoulli. Then for all  $k$  sufficiently large we have  $E_\nu \bar{f}_k(\nu^{X^k}, \nu^{X^k}(\cdot | X_{-\infty}^0)) < 84\varepsilon/(1 - \varepsilon)$ .

PROOF. By inequality (1), with  $Y$  a trivial process, all we need to is find some  $k$  for which  $E_\nu \bar{f}_k(\nu^{X^k}, \nu^{X^k}(\cdot | X_{-\infty}^0)) < 42\varepsilon/(1 - \varepsilon)$ .

By [7, proposition 2.6] there exists a probability space  $(\Omega, \mathcal{F}, P)$  and processes  $\tilde{U}, \tilde{V}, U, V, M_U, M_V$  defined on it such that the following conditions hold:

(a)  $\tilde{U}, \tilde{V}$  have state space  $\hat{B}$ ;  $U, V$  have state space  $B$  and are obtained from  $\tilde{U}, \tilde{V}$  respectively, by concatenation as in Lemma 4;  $M_U, M_V$  have state space  $\{0, 1\}$  and  $(M_U)_i = 1$  if and only if  $U_i$  is the left-most entry of the  $\tilde{U}_i$  in which it lies; similarly for  $M_V$ , vis-a-vis  $V, \tilde{V}$ ,

(b)  $\tilde{U}, \tilde{V}$  are jointly stationary;  $|P^V - \nu| < 2\varepsilon$ ; and  $|P^{(U, M_U)} - \lambda| < 2\varepsilon$ , for some stationary  $\lambda$  on  $B^\infty \times \{0, 1\}^\infty$  with  $B^\infty$ -marginal  $\mu$ ,

(c)  $EI(\tilde{U}_0) = EI(\tilde{V}_0) < 1/(1 - \varepsilon)$ , and

(d) with probability one, for each  $i$ , the outputs  $\tilde{U}_i, \tilde{V}_i$  begin with the same element of  $B$ .

Let  $\bar{X}, \bar{Y}$  be the processes which are the projections from  $B^\infty \times \{0, 1\}^\infty \rightarrow B^\infty$ ,

$\{0, 1\}^\infty$  respectively. Let  $\tau$  represent the joint distribution  $\tau = P^{(U, M_U)}$ . By Lemma 2, for any  $m$ ,

$$\begin{aligned} E\bar{f}_m(P^{U^m}, P^{U^m}(\cdot | U_{-\infty}^0, (M_U)_{-\infty}^0)) &= E\bar{f}_m(\tau^{\bar{X}^m}, \tau^{\bar{X}^m}(\cdot | \bar{X}_{-\infty}^0, \bar{Y}_{-\infty}^0)) \\ &\leq E\bar{f}_m(\lambda^{\bar{X}^m}, \lambda^{\bar{X}^m}(\cdot | \bar{X}_{-\infty}^0, \bar{Y}_{-\infty}^0)) + 6|\tau - \lambda|. \end{aligned}$$

From condition (b) above,  $|\tau - \lambda| < 2\varepsilon$ , so the second term on the right is bounded by  $12\varepsilon$ . As  $m$  gets large, the first term goes to zero by Lemma 3. (To see this, note that since  $B$  is finite, the second term on the right side of (1) is identically zero. The first term on the right side of (1) goes to zero since  $\mu$  is LB.) Hence we see that we may fix  $m$  so that

$$(2) \quad E\bar{f}_m(P^{U^m}, P^{U^m}(\cdot | U_{-\infty}^0, (M_U)_{-\infty}^0)) < 12\varepsilon.$$

Now Lemma 4 yields that

$$E\hat{f}_m(P^{\hat{U}^m}, P^{\hat{U}^m}(\cdot | \hat{U}_{-\infty}^0)) \leq E\bar{f}_m(P^{U^m}, P^{U^m}(\cdot | \hat{U}_{-\infty}^0)) + \varepsilon/(1 - \varepsilon),$$

since by hypothesis (c)  $El(\hat{U}) < 1/(1 - \varepsilon)$ . Since the first term is bounded by  $12\varepsilon$  from (2), we conclude that

$$(3) \quad E\hat{f}_m(P^{\hat{U}^m}, P^{\hat{U}^m}(\cdot | \hat{U}_{-\infty}^0)) < 13\varepsilon/(1 - \varepsilon).$$

Since the entropies  $H(\hat{U}_0)$  and  $H(\hat{V}_0)$  are finite [4, theorem 4] and since  $(\hat{U}, \hat{V})$  are jointly stationary, theorem 6.2 [6, p. 66] assures us that the conditional mutual information  $I(U^m, \hat{V}_{-\infty}^{-i} | \hat{U}_{-\infty}^0) \rightarrow 0$  as  $i$  gets large. Hence

$$\limsup_{i \rightarrow 0} E\bar{f}_m(P^{U^m}(\cdot | \hat{U}_{-\infty}^0), P^{U^m}(\cdot | \hat{U}_{-\infty}^0, \hat{V}_{-\infty}^{-i})) = 0$$

by exactly the same argument given in the proof of theorem 1 of [3]. Thus since  $El(\hat{U}_0) - 1 < \varepsilon/(1 - \varepsilon)$  by condition (c) above, it follows immediately from Lemma 4 that

$$(4) \quad \limsup_{i \rightarrow \infty} E\hat{f}_m(P^{\hat{U}^m}(\cdot | \hat{U}_{-\infty}^0), P^{\hat{U}^m}(\cdot | \hat{U}_{-\infty}^0, \hat{V}_{-\infty}^{-i})) < \varepsilon/(1 - \varepsilon).$$

Therefore, by reapplying Lemma 3 for  $k$  any sufficiently large multiple of  $m$ , we get that

$$(5) \quad E\hat{f}_k(P^{\hat{U}^k}, P^{\hat{U}^k}(\cdot | \hat{U}_{-\infty}^0, \hat{V}_{-\infty}^0)) < 27\varepsilon/(1 - \varepsilon).$$

(The terms on the right-hand side of Equation (1) are bounded above in (3) and (4).)

On the other hand, both  $E\hat{f}_k(P^{\hat{U}^k}(\cdot | \hat{U}_{-\infty}^0, \hat{V}_{-\infty}^0), P^{\hat{V}^k}(\cdot | \hat{U}_{-\infty}^0, \hat{V}_{-\infty}^0))$  and

$E\hat{f}_k(P^{\tilde{U}^k}, P^{\tilde{V}^k})$  are dominated by  $E\hat{f}_k(\tilde{U}^k, \tilde{V}^k) \leq E[(l(\tilde{U}^k) + l(\tilde{V}^k) - 2k)/(2k)] < \varepsilon/(1 - \varepsilon)$ . From these two bounds and (5), the triangle inequality yields that

$$(6) \quad E\hat{f}_k(P^{\tilde{V}^k}, P^{\tilde{V}^k}(\cdot | \tilde{U}_{-\infty}^0, \tilde{V}_{-\infty}^0)) < 29\varepsilon/(1 - \varepsilon).$$

Since  $E\bar{f}_k(P^{\tilde{V}^k}, P^{\tilde{V}^k}(\cdot | V_{-\infty}^0)) \leq E\bar{f}_k(P^{\tilde{V}^k}, P^{\tilde{V}^k}(\cdot | \tilde{V}_{-\infty}^0))$  and since Lemma 4 yields  $E\bar{f}_k(P^{\tilde{V}^k}, P^{\tilde{V}^k}(\cdot | \tilde{V}_{-\infty}^0)) \leq E\hat{f}_k(P^{\tilde{V}^k}, P^{\tilde{V}^k}(\cdot | \tilde{V}_{-\infty}^0)) + \varepsilon/(1 - \varepsilon)$ , we apply (6) to conclude that

$$(7) \quad E\bar{f}_k(P^{\tilde{V}^k}, P^{\tilde{V}^k}(\cdot | V_{-\infty}^0)) < 30\varepsilon/(1 - \varepsilon).$$

Finally, by Lemma 2,

$$\begin{aligned} E\nu\bar{f}_k(\nu^{X^k}, \nu^{X^k}(\cdot | X_{-\infty}^0)) &\leq E\bar{f}_k(P^{\tilde{V}^k}, P^{\tilde{V}^k}(\cdot | V_{-\infty}^0)) + 6|P^V - \nu| \\ &\leq 42\varepsilon/(1 - \varepsilon), \end{aligned}$$

since the term  $|P^V - \nu|$  is less than  $2\varepsilon$  by condition (b) above and the previous term is bounded in (7).

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