Time Shifted Pilot Signal Transmission With Pilot Hopping To Improve The Uplink Performance of Massive MIMO System For Next Generation Network

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Abstract

The paucity of pilot signals in Massive MIMO system is a vital issue. To accommodate substantial number of users, pilot signals are reused. This leads to interference, resulting in pilot contamination and degrades channel estimation at the Base Station (BS). Hence, mitigation of pilot contamination is exigency in Massive MIMO system. The proposed Time Shifted Pilot Signal Transmission with Pilot signal Hopping (TSPTPH), addresses the pilot contamination issue by transmitting pilot signals in non-overlapping time interval with hopping of pilot signals in each transmission slot. Hopping is carried by switching user to a new pilot signal in each transmission slot, resulting in random change of interfering users. This contributes to the change in channel coefficient, which leads to improved channel estimation at the BS and therefore enhances the efficiency of Massive MIMO system. In this system, Uplink Signal Power to Interference plus Noise Power Ratio (SINR) and data-rate are calculated for pilot signal reuse factor 1 and 3, by estimating the channel with Least Square estimation. The proposed system also reduces the uplink Signal power for data transmission of each User Equipment for normalized spectral efficiency with rising number of antennas at the BS and thus improves battery life.

Keywords: Channel estimation, Hopping, Pilot contamination, Pilot signal, Signal Power to Interference plus Noise Power Ratio.
1. Introduction

In the next generation network, desire for high data-rate and reliability is increasing considerably, whilst at the same time user density and traffic intensity are increasing exponentially. Therefore, the design of the cellular network is requisite to provide a reliable and good quality of service. To a large extent, this can be taken care of by deploying a significant number of $M$ antennas at the BS, providing services to $K$ number of User Equipment (UE). Main merits of using a vast number of antennas at the BS are mentioned by [1], [2] and [3]. It not only provides reliability and quality of service but also has the following advantages:

- It reduces the radiated power of each UE to $1/M$ for perfect Channel State information (CSI) and $1/\sqrt{M}$ for imperfect CSI thereby, increasing the battery life.
- Channel vector is random between UE and BS, and becomes pairwise orthogonal with $M$.
- With simple signal processing, uncorrelated noise and intra-cell interference can be eliminated, and the effect of small-scale fading and thermal noise can be averaged out. Basics of MMIMO system with emphases on the challenges and opportunities are explained in [4]. In Massive MIMO (MMIMO) system channel coefficient matrix for $K$ UEs in the $l^{th}$ cell to the $n^{th}$ BS can be expressed as

$$H_{nl} = D_{nl}^{1/2}G_{nl}$$

where, $H_{nl} \in \mathbb{C}^{M \times K}$ and is a combination of small-scale and large-scale fading coefficients, $G$ is a $M \times K$ matrix of small-scale fading coefficients, $D$ is a $K \times K$ diagonal matrix whose diagonal elements comprise the vector $[\beta_1, \beta_2, \ldots, \beta_M]$, and $D_{nl} = \text{diag}[\beta_{n1}, \beta_{n2}, \ldots, \beta_{nk}]$. The Rayleigh channel vector between $k^{th}$ user in the $l^{th}$ cell to the $i^{th}$ antenna of the $n^{th}$ BS is represented by $h_{nli}$ and is given as

$$h_{nli} = \sqrt{\beta_{nli}}g_{nli}$$

here, $g_{nli}$ is a $1 \times M$ small-scale fading vector with $\sim \mathcal{CN}(0, I_M)$ unknown random variables. $\beta_{nli}$ represents large-scale fading coefficient (geometric and shadow fading), and it is a function of $z_{nli}$ and radial distance $r_{nli}$ and given as

$$\beta_{nli} = \frac{z_{nli}}{r_{nli}^{\gamma}}$$

Cells of same size may have different environments due to shadowing. Therefore, $z_{nli}$ is a random variable having lognormal distribution about mean and changes slowly over space [5]. The quantity $10\log_{10}(z_{nli})$ is Gaussian with zero mean and standard deviation of $\sigma_{\text{shad}}$. The value of $\sigma_{\text{shad}}$ varies from $6dB$ to $12dB$ or even higher in dense environment and $r_{nli}$ is the radial distance between $k^{th}$ UE in the $l^{th}$ cell and the BS in the $n^{th}$ cell, $\gamma$ is the propagation path loss exponent. Therefore statistically, parameters which define the path loss due to
large-scale fading for dynamically located UE are radial distances between interfering UEs and BS, standard deviation $\sigma$ and path loss exponent $\gamma$.

2. Related Work

Pilot contamination is alarming if same set of the pilot sequences is reused in multi-cell MMIMO system. Vidit et al. [6] have suggested a scheme which reduces the pilot contamination and improves the downlink SINR and sum-rate by reusing pilot signals with power control. A smart pilot assignment scheme has been proposed by Zhu et al. [7], in this scheme pilot sequence is allocated to each user after analysing the channel quality (large-scale fading coefficient) and quantifying the inter-cell interference. Researchers [8], [9] and [10] have used linear precoding techniques such as Matched Filtering (MF), Zero Forcing (ZF) and Minimum Mean Square Error (MMSE) to improve system performance such as uplink and downlink SIR and spectral efficiency in the presence of pilot contamination. Improvement in downlink performance of MMIMO system using fractional frequency reuse with MF and ZF precoding is demonstrated in our previous work [11]. Improvement in achievable data-rate when the number of antennas approaches to infinity with irregular antenna array installation is shown in [12]. Pilot contamination is severe in multi-cell MMIMO system, where users transmit pilot sequence synchronously. Kumar et al. [13] have proposed a model in which pilot are transmitted asynchronously. It reduces inter-cell interference and improves channel estimation. Time-shifted pilot sequence transmission with covariance based channel estimation scheme has been proposed by Yao et al. [14] to improve the downlink performance. To improve the performance of MMIMO system, the concept of asynchronous pilot transmission is further carried by other researchers [15], [16], [17] and [18] by using MMSE channel estimation and linear pre-coding techniques such as Matched Filtering and Zero Forcing. Jesper et al. [19] introduced a novel scheme in which users transmit pilot sequence synchronously and each user hops to new pilot sequence in each transmission slot. To improve the performance of MMIMO system in the presence of pilot contamination, a scheme is proposed in our previous work [20], where pilot signals are transmitted asynchronously, and pilot signal hopping is performed in each transmission slot. The change in interfering users in each transmission slot due to hopping, causes change in channel coefficient which results in better channel estimation. In this proposed work, we now go ahead to improve the uplink performance of MMIMO system.

In the proposed TSPTPH system, pilot signal is transmitted by UEs asynchronously with pilot hopping at each transmission slot. Due to hopping, the group of UEs contaminating the channel will change in each transmission slot and therefore improves channel estimation. In this system, pilot signal transmission is overlapped with the downlink data and channel is estimated using Least Square estimation. The results are analysed and compared with Time-Shifted Pilot signal Transmission (TSPT) [16] system and shows that there is significant improvement in uplink performance.

Notation: Matrices are represented by upper case boldface, vectors are represented by lower case boldface, $(\cdot)^T$ is matrix transpose, $(\cdot)^H$ is Hermitian transpose and $x \sim CN(a;b)$ indicates variable $x$ has a circularly symmetric complex Gaussian distribution with mean $a$ and variance $b$. 

\[
\begin{align*}
\text{large-scale fading for dynamically located UE are radial distances between interfering UEs and BS, standard deviation } \sigma \text{ and path loss exponent } \gamma. \\
2. \text{Related Work} \\
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3. Proposed TSPTPH system model

A multi-cell network with $L$ hexagonal cells is considered where, each cell is provided with centrally located BS having vast number of $M$ omni-directional antennas. Each cell is having $K$ users, each with single antenna and are distributed randomly in a cell. For pilot signal reuse factor $1/N$ where, $N = 1$ and $3$, there are six interfering cells. Fig. 1 Shows seven cells structure for $N = 1$, and the intended UE is in the centre cell $A$ and there are six interfering cells. For $N = 3$, each cluster will have 3 cells with similar configuration. In the proposed system, the pilot signal transmission is overlapped with the downlink data and simultaneously pilot signal hopping is performed in each transmission slot.

Fig. 1. Group of users contaminating the channel

Fig. 2. Time shifted pilot sequence transmission

$B_1$, where UEs transmit pilot signals to their respective BS at the beginning of every coherence interval. Coherence interval of $B_2$ group consists of cells $B, D, F$ and starts with downlink data transmission as shown in Fig. 2. An orthogonal time-frequency pilot signal assigned to each UE within a cell. Pilot signals used by $K$ UEs can be represented by $\tau \times K$ matrix and $\psi = (\psi_1, \psi_2, \ldots, \psi_K)$ where, $(\tau \geq K)$ and $k^{th}$ UE of all the cells transmits the same pilot signal $\psi_K$. Pilot signals are normalized such that $\psi_k^H \psi_k = K \delta_{k,k}$. TDD protocol is used to get the CSI and channel vector is considered to be flat during transmission slot of T symbols period. In each transmission slot, four operations are performed:
- Training phase: Each UE sends orthogonal pilot signal to the respective BS for $N_p$ symbol period
- Channel estimation: On receiving pilot signal, BS estimates channel vector for the corresponding user in $N_c$ symbol period
- Uplink data transmission: Users send uplink data to its BS for $N_u$ symbol period
- Downlink data transmission: BS transmits downlink data to the corresponding users for $N_d$ symbol period

$$T = N_p + N_c + N_u + N_d$$ (4)

An unique pilot signal hopping pattern which is known to BS is assigned to each UE within a cell as shown in Table 1. UEs switch to a new pilot signal in each transmission slot, this causes random change in group of interfering users. Therefore, the channel coefficient i.e. the large-scale fading coefficient $\beta$ which is a function of $z$ (log-normal distributed random variable) and $r$ (radial distance between UE and BS) will also change in every transmission slot as depicted in Fig. 1. Thus, in each transmission slot, channel is contaminated by dynamically changing interfering users due to pilot hopping. This randomization of the interfering users significantly reduces the channel estimation error. Whereas, in conventional MMIMO system and TSPT system channel coefficients between UEs and BS will remain same since there is no change in the interfering users.

Table 1. Pilot sequence hopping pattern

| Users | T1 | T2 | T3 | T4 | T5 | T6 |
|-------|----|----|----|----|----|----|
| 01    | $\Psi_1$ | $\Psi_5$ | $\Psi_4$ | $\Psi_3$ | $\Psi_2$ | $\Psi_6$ |
| 02    | $\Psi_3$ | $\Psi_2$ | $\Psi_6$ | $\Psi_5$ | $\Psi_4$ | $\Psi_1$ |
| 03    | $\Psi_4$ | $\Psi_6$ | $\Psi_5$ | $\Psi_2$ | $\Psi_1$ | $\Psi_3$ |
| 04    | $\Psi_5$ | $\Psi_4$ | $\Psi_3$ | $\Psi_6$ | $\Psi_2$ | $\Psi_1$ |
| 05    | $\Psi_2$ | $\Psi_3$ | $\Psi_1$ | $\Psi_4$ | $\Psi_6$ | $\Psi_4$ |

3.1 Training Phase

In training phase, UEs of group $B_1$ cells transmit the pilot signal, and BSs of group $B_2$ cells transmit downlink data at the beginning of each coherence interval. The pilot signal received at the BS $A$ is $\mathbf{Y}_A$ given as

$$\mathbf{Y}_A = \sum_{i \in B_1} \sum_{k=1}^K \sqrt{\rho_p} \sqrt{\beta_{Ai}} \mathbf{w}_k \mathbf{g}_{Ai} + \sum_{m \in B_2} \sum_{k=1}^K \sqrt{\rho_d} \mathbf{q}_{mk} \mathbf{w}_m \mathbf{J}_{Am} + \mathbf{V}_A$$ (5)

Where, $\mathbf{Y}_A$ is $K \times M$ received signal matrix at the BS $A$, $\beta_{Ai}$ and $\mathbf{g}_{Ai}$ are large-scale and small-scale fading coefficient between $k^{th}$ user of $i^{th}$ cell and $A^{th}$ BS, $\mathbf{q}_{mk}$ and $\mathbf{J}_{Am}$ are large-scale and small-scale fading coefficient between the BS $A$ and the BS $m$, and $\mathbf{J}_{Am} \sim CN(0, \mathbf{I}_{M \times M})$. Since, BSs are stationary, the channel characteristic between BSs
change very slowly. \( \psi_k \) is \( K \times 1 \) pilot signal vector of \( k \)th UE, \( \rho_p \) is the measure of pilot signal power to noise variance ratio and \( \rho_d \) is the ratio of downlink data power to noise variance. Here, noise variance is considered to be one, \( \mathbf{q}_{mk} \) is \( K \times 1 \) message bearing symbol for the \( k \)th UE from the \( m \)th BS and \( \mathbf{q}_{mk} \sim CN\left(0, \mathbf{I}_K\right) \). It means, symbol vectors are uncorrelated and having variance one i.e. \( E\left\{\mathbf{q}_{mk}\mathbf{q}_{mk}^H\right\} = \mathbf{I}_K \), \( \mathbf{V}_A \) is \( K \times M \) additive noise matrix, which is defined as IID, \( CN\left(0, \mathbf{I}_{KM}\right) \) and \( \mathbf{w}_{mk} \) is \( M \times 1 \) pre-coding vectors for the \( k \)th user from the \( m \)th BS.

### 3.2 Channel Estimation

On receiving the pilot signals, the \( A \)th BS estimates the channel vector for its \( k \)th UE using the Least Square estimation by correlating it with known pilot signal and is given by

\[
\hat{\mathbf{g}}_{A,k} = \psi_k^H \times \mathbf{Y}_A
\]

\[
\beta_{A,k} = \sqrt{\rho_p} \sum_{i \in B_1} \sum_{k=1}^K \beta_{Aik} \mathbf{g}_{Aik}, K + \sum_{m \in B_2} \sum_{k=1}^K \rho_d \mathbf{A}_{Amk} \psi_k^H \mathbf{J}_{Amk} \mathbf{w}_{mk} \mathbf{q}_{mk} + \psi_k^H \mathbf{V}_A
\]

The Estimated channel for \( k \)th UE is given in equation 7, it shows that channel is contaminated by the pilot signals transmitted by UEs of group \( B_1 \) cells (first term), downlink data transmitted by BSs belonging to group \( B_2 \) (second term) and the noise. The first term \( \beta_{Aik} \) is the function of radial distance between UE and BS and will vary in each transmission slot as the group of UEs changes randomly due to pilot signal hopping. The covariance of the estimated channel is calculated by considering that, every random variable in equation 7 is uncorrelated and therefore covariance is given as

\[
C_{\hat{\mathbf{g}}_{A,k}, \hat{\mathbf{g}}_{A,k}} = \sum_{i \in B_1} \sum_{k=1}^K \rho_p \beta_{Aik}^2 \mathbf{I}_M + \sum_{m \in B_2} \sum_{k=1}^K \rho_d \lambda_{Amk} K \mathbf{I}_M + \phi_{A,k} \mathbf{I}_M
\]

where,

\[
\phi_{A,k} = \sum_{i \in B_1} \rho_p \beta_{Aik} - \sum_{m \in B_2} \sum_{k=1}^K \rho_d \lambda_{Amk} + 1
\]

The estimation error vector \( \hat{\mathbf{g}}_{A,k} = \hat{\mathbf{g}}_{A,k} - \mathbf{g}_{A,k} \), here \( \hat{\mathbf{g}}_{A,k} \) and \( \mathbf{g}_{A,k} \) are uncorrelated therefore the covariance matrix of \( C_{\hat{\mathbf{g}}_{A,k}, \hat{\mathbf{g}}_{A,k}} \) is given as \( (1-\phi_{A,k}) \mathbf{I}_M \). It is considered that estimated channel is having Gaussian distribution, thus \( \hat{\mathbf{g}}_{A,k} \) and \( \mathbf{g}_{A,k} \) have distribution \( CN\left(0, (1-\phi_{A,k}) \mathbf{I}_M\right) \) and \( CN\left(0, \mathbf{I}_M\right) \) respectively.

### 3.3 Uplink data transmission

All the UEs of group \( B_1 \) and \( B_2 \) cells transmit uplink data simultaneously to their respective BSs. And BSs receive \( M \times 1 \) vectors of message symbol from each UE. Therefore \( A \)th BS receives the uplink data transmitted by its \( k \)th UE is given as
\[ d_{Ak} = \sum_{i=1}^{K} \sqrt{\rho_i} \sqrt{\beta_{Ak}} g_{Ak} d_{ik} + \sum_{m=1}^{K} \sum_{k=1}^{K} \sqrt{\rho_m} \sqrt{\beta_{mk}} g_{mk} d_{mk} + n_A \]  

(11)

Here, \( n_A \) is a \( 1 \times M \) additive noise vector between the BS \( A \) and the \( k \)-th user with entries IID, \( CN(0, I_M) \), \( \rho_A \) is the ratio of uplink data transmission power to noise variance and \( d_{ik} \) is the uplink data transmitted from \( k \)-th UE. Maximum Ratio Combining (MRC) is used at the BS to calculate the estimated data which is a function of \( \hat{g}_{Ak} \). Therefore, estimated data for the \( k \)-th user of cell \( A \) is given as

\[ d_{Ak^*} = d_{Ak} f_{Ak^*} \]  

(12)

where, \( f_{Ak^*} \) is the beam forming vector and it is defined as

\[ f_{Ak^*} = \frac{g_{Ak^*}^H}{K \|g_{Ak^*}\|} \]  

(13)

From equations. 11 and 12 the estimated data for \( k \)-th user is obtained as

\[ d_{Ak^*} = \sum_{i=1}^{K} \sqrt{\rho_i} \sqrt{\beta_{Ak^*}} g_{Ak^*} d_{ik} + \sum_{m=1}^{K} \sum_{k=1}^{K} \sqrt{\rho_m} \sqrt{\beta_{mk}} g_{mk} d_{mk} f_{Ak^*} + n_A f_{Ak^*} \]  

(14)

SINR for the \( k \)-th UE of cell \( A \) is defined as \( \frac{S}{N_a} \). Signal power \( S_a \) is illustrated as

\[ S_a = \rho_A \beta_{Ak^*} |E[g_{Ak^*} f_{Ak^*}]|^2 \]  

(15)

The achievable uplink data rate for the \( k \)-th UE of cell \( A \) is given as

\[ R_{Ak^*}^{Up} = \frac{N_u}{T} \log_2 \left( 1 + \text{SINR}_{Ak^*}^{Up} \right) \]  

(16)

where, \( N_u \) is the uplink data symbol period. The Complexity of expectation in equation. 16 is reduced by using Jensen’s inequality [16] to calculate estimated achievable uplink data-rate and is given as

\[ R_{Ak^*}^{Up} \geq \overline{R}_{Ak^*}^{Up} \]  

(17)

\[ \overline{R}_{Ak^*}^{Up} = \frac{N_u}{T} \log_2 \left( \left( E \left[ \frac{N_a}{S_a} \right] \right)^{-1} \right) \]  

(18)

Therefore, \( S_a \) can be expressed as

\[ S_a = \rho_A K \beta_{Ak^*} (M - 1) \]  

(19)

And the expected value of \( \frac{N_a}{S_a} \) can be expanded as

\[ E \left[ \frac{N_a}{S_a} \right] = \left[ \rho_A \beta_{Ak^*} E \left[ \frac{g_{Ak^*} f_{Ak^*}^2}{S_a} \right] + \sum_{i \neq k, k \neq l} \rho_A \beta_{Ak^*} E \left[ \frac{g_{Ak^*} f_{Ak}^2}{S_a} \right] \right] \]

(20)
By using equations 19 and 20, the closed form expression for Uplink SINR is inferred as

\[
SINR_{Auk}^{Up} = \frac{\rho_u K \beta_{Auk} \phi_{Auk} \left( M - 1 \right)}{\rho_u \beta_{Auk} \left( 1 - \phi_{Auk} K \right) + \sum_{i\neq A, i\in L} \rho_u \beta_{Aik} \left( \phi_{ik} K (M - 1) + 1 \right) + \sum_{i\in L} \sum_{k \neq k'} \rho_u \beta_{Aik} + 1}
\]

(21)

The expression for achievable uplink data-rate for the \( k^{th} \) UE of cell \( A \) is given as

\[
\bar{R}_{Auk}^{Up} = \frac{N_u}{T} \log_2 \left( 1 + \frac{\rho_u K \beta_{Auk} \phi_{Auk} \left( M - 1 \right)}{\rho_u \beta_{Auk} \left( 1 - \phi_{Auk} K \right) + \sum_{i\neq A, i\in L} \rho_u \beta_{Aik} \left( \phi_{ik} K (M - 1) + 1 \right) + \sum_{i\in L} \sum_{k \neq k'} \rho_u \beta_{Aik} + 1} \right)
\]

(22)

In TSPT, interfering users remain same as there is no hopping whereas in TSPTPH, due pilot signal hopping, interfering users change in each transmission slot. From equation. 21 and 22, it is observed that SINR and data-rate are function of \( \beta_{Aik} \), \( K \) and \( M \), first term of the denominator is the channel estimation error and second term is the interference caused by UEs of group \( B_1 \) and \( B_2 \) cells transmitting the same pilot signal except cell \( A \). And, these interfering UEs change dynamically in each transmission slot due to the pilot signal hopping. Therefore, \( \beta_{Aik} \) and \( \phi_{ik} \) are function of radial distance between UEs and BS also change as explained (see Appendix). Third term is the interference caused by users of group \( B_1 \) and \( B_2 \), where \( k \neq k' \) which also varies in each transmission slot due to pilot hopping and fourth term is the noise variance.

5. Simulation Results and analysis

The proposed TSPTPH system is evaluated for \( K = 8 \) and \( K = 15 \), pilot signal hopping is done for 30 and 40 transmission slots and results are compared with TSPT for reuse \( N = 8 \) and \( N = 3 \), by varying the BS antennas from 10 to 250. Simulations are carried out in MATLAB for 1500 trials and average values are finally used for analysis of the results. Cells are considered to be of the same size with radius of 1.5 km and having no users within 100 m radius of cell. Each cell is provided with equal number of UEs that are distributed randomly. \( \rho_p = 10 \text{dB} \), \( \rho_u = 10 \text{dB} \) and \( \rho_d = 20 \text{dB} \) are the measures of SNR during pilot signal, uplink and downlink data transmission phase respectively. The noise variance is considered to be one and all the transmission slots have \( T = 17 \), \( N_u = 6 \), \( N_d = 6 \), \( N_c = 2 \), \( N_p = 3 \).
In mobile system, propagation environment of cells may have different characteristics though they are of the same size. Therefore, it is required to calculate variation about mean. In [21], path loss $L_p(d)$ is a random variable having log-normal distribution about the mean, which is distant-dependent value $\overline{L}_p(d)$. Thus $L_p(d)$ can be expressed as

$$L_p(d)(dB) = L_i(d_o)(dB) + 10\gamma \log\left(\frac{r}{d_o}\right) + X_\sigma(dB)$$

(23)

Here, $X_\sigma$ is zero-mean Gaussian random variable with standard deviation $\sigma_{\text{shad}}(dB)$. $X_\sigma$ is equivalent to $z$ and dependent on cell environment and cell size. Thus, analysis of the Rayleigh fading channel is carried out for three path loss models taking into consideration shadow fading $\sigma_{\text{shad}}(dB) = 0dB, 8dB$ and $12dB$ with path-loss exponent $\gamma = 3.8$.

The pseudo code of the implementation is given below Algorithm 1. Uplink SINR and data-rate for the three path loss models for $K=8$ with 40 hops and 30 hops for reuse factor $N=1$ and $N=3$, are shown in Fig. 3 and Fig. 4. The corresponding observations are inferred in Table 2 - Table 5. The analysis of the observations is given below.

**Reuse factor $N=1$**
- Performance of TSPTPH is slightly better for $M<110$ both for 30 hops and 40 hops for $\sigma_{\text{shad}} = 0dB$. It starts decreasing slowly for 30 hops when $M>200$ for both TSPTPH and TSPT. For $\sigma_{\text{shad}} = 8dB$, both systems perform nearly same. As the effect of fading increases i.e. $\sigma_{\text{shad}} = 12dB$, TSPTPH performs much better and shows average improvement of 4.2% for 40 hops. For 30 hops, TSPT performs better for $M>150$.

**Reuse factor $N=3$**
- TSPTPH performs significantly better till $M<220$ for $\sigma_{\text{shad}} = 0dB$ with 40 hops and shows 3.4% improvement and it is 1.8% for 30 hops when $M>90$. For $\sigma_{\text{shad}} = 8dB$, performance of TSPTPH much better than TSPT and increases continuously with M and shows 3% improvement for 40 hops, whereas for 30 hops TSPTPH performs well till $M<100$ and then TSPT performs good and shows improvement of 2.5% with respect to TSPTPH. For $\sigma_{\text{shad}} = 12dB$, TSPTPH performs much better and shows improvement of 4% over TSPT for 40 hops whereas for 30 hops, performance of TSPT is better for $M<130$ and then after TSPTPH shows improvement of 1.5%.

### Table 2. Uplink SINR (dB) $K=8$, $N=1$

| Path-loss model $\sigma_{\text{shad}}$ | Number of hops = 40 | Number of hops = 30 |
|--------------------------------------|---------------------|---------------------|
| | $M=100$ | $M=200$ | $M=100$ | $M=200$ | $M=100$ | $M=200$ |
| 0dB | 39.1 | 39.4 | 43 | 41.8 | 39.3 | 40.7 | 43.9 | 43.3 |
| 8dB | 40.25 | 39.9 | 44 | 43.5 | 42.1 | 41.6 | 44.9 | 43.8 |
| 12dB | 38.8 | 41.8 | 43.1 | 45 | 43.7 | 40 | 45.4 | 43.5 |
Similarly Uplink SINR and data-rate for the three path-loss models for reuse factor $N=1$ and 3

**Algorithm 1.** Pseudo Code of the implementation

Assign pilot sequence as shown in Table I
Select number of hops
Select number of iterations
for reuse factor $N=1$ and 3
    for iteration = 1 to number of iterations
        for transmission slots = 1 to number of hops
            for antennas = 1 to 250
                Calculate channel estimation
                Calculate Beamforming vector
                Calculate Uplink SINR
                Calculate Uplink achievable data-rate
                Calculate Uplink transmission power (SNR)
            end
        end
    end
Average uplink SINR for iterations
Average uplink data-rate for iterations
Average uplink transmission power for iterations

Similarly Uplink SINR and data-rate for the three path-loss models for reuse factor $N=1$
and \( N = 3 \), for \( K = 15 \) with 40 hops and 30 hops are shown in Fig. 5 and Fig. 6. The corresponding observations are depicted in Table 6 - 9 respectively. Observations are

- Performance of TSPT is better for \( M > 190 \) and then TSPTPH start improving for 40 hops with \( \sigma_{\text{shad}} = 0 dB \). For \( \sigma_{\text{shad}} = 8 dB \), TSPTPH performs significantly better and shows 5% improvement for 40 hops and 4% for 30 hops over TSPT. In severe fading environment i.e. \( \sigma_{\text{shad}} = 12 dB \), TSPTPH gives 3% improvement over TSPT for 40 hops and for 30 hops TSPT performance improves by 6%.

- TSPTPH performs much better and improves performance by 4.5% for 40 hops with \( \sigma_{\text{shad}} = 0 dB \), whereas for 30 hops TSPTPH shows 3% improvement for \( M < 140 \). For \( \sigma_{\text{shad}} = 8 dB \), performance of TSPTPH shows 2% for 40 hops and 3.5% for 30 hops improvement over TSPT. For \( \sigma_{\text{shad}} = 12 dB \), TSPTPH gives 2% improvement for 30 hops and for 40 hops it performs well for \( M > 110 \) with 3% improvement over TSPT.

### Table 6. Uplink SINR (dB) \( K = 15, N = 1 \)

| Path-loss model | Number of hops = 40 | Number of hops = 30 |
|-----------------|---------------------|---------------------|
| \( \sigma_{\text{shad}} \) | \( M=100 \) | \( M=200 \) | \( M=100 \) | \( M=200 \) |
|-----------------|---------------------|---------------------|
| 0dB             | TSPT | TSPTPH | TSPT | TSPTPH | 39.7 | 42 | 45.1 | 42.25 |
| 8dB             | 41 | 42.5 | 44.5 | 47 | 41 | 42.9 | 43.8 | 45.6 |
| 12dB            | 39.4 | 39.8 | 44 | 46.5 | 41 | 37 | 48 | 46.2 |

### Table 7. Uplink data-rate (bps / Hz) \( K = 15, N = 1 \)

| Path-loss model | Number of hops = 40 | Number of hops = 30 |
|-----------------|---------------------|---------------------|
| \( \sigma_{\text{shad}} \) | \( M=100 \) | \( M=200 \) | \( M=100 \) | \( M=200 \) |
|-----------------|---------------------|---------------------|
| 0dB             | TSPT | TSPTPH | TSPT | TSPTPH | 5.1 | 4.85 | 5.22 | 5.24 |
| 8dB             | 4.8 | 5.0 | 5.2 | 5.5 | 4.8 | 5.03 | 5.13 | 5.35 |
| 12dB            | 4.62 | 4.66 | 5.18 | 5.45 | 4.8 | 4.37 | 5.6 | 5.4 |

### Table 8. Uplink SINR (dB) \( K = 15, N = 3 \)

| Path-loss model | Number of hops = 40 | Number of hops = 30 |
|-----------------|---------------------|---------------------|
| \( \sigma_{\text{shad}} \) | \( M=100 \) | \( M=200 \) | \( M=100 \) | \( M=200 \) |
|-----------------|---------------------|---------------------|
| 0dB             | TSPT | TSPTPH | TSPT | TSPTPH | 46.3 | 48.6 | 52 | 54 |
| 8dB             | 46.4 | 48 | 51.25 | 51.4 | 45.5 | 46.5 | 54.8 | 57.2 |
| 12dB            | 46.4 | 46.2 | 54.8 | 56.3 | 42.7 | 43.8 | 50.5 | 51.7 |

### Table 9. Uplink data-rate (bps / Hz) \( K = 15, N = 3 \)

| Path-loss model | Number of hops = 40 | Number of hops = 30 |
|-----------------|---------------------|---------------------|
| \( \sigma_{\text{shad}} \) | \( M=100 \) | \( M=200 \) | \( M=100 \) | \( M=200 \) |
|-----------------|---------------------|---------------------|
| 0dB             | TSPT | TSPTPH | TSPT | TSPTPH | 5.43 | 5.7 | 6.1 | 6.32 |
| 8dB             | 5.43 | 5.64 | 6.01 | 6.025 | 5.33 | 5.45 | 6.4 | 6.7 |
| 12dB            | 5.44 | 5.42 | 6.4 | 6.6 | 5.01 | 5.14 | 5.96 | 6.13 |
From the above analysis it is observed that performance of uplink is better for $N=3$. In this case, the interfering cells are far apart i.e. $\sqrt{3N}$ from the central cell, therefore, distance between the UE and the interfering cells UEs is very large.
Thus, it reduces the denominator term $\beta_{sk}$ of equations 10, 21 and 22 which results in increase in SINR and thus data-rate. Further, the analysis of uplink data transmission power required per UE for imperfect CSI to achieve spectral efficiency of 1bit/sec/Hz with respect to BS antennas for $K=15$, is depicted in Fig. 7. It shows that, as the number of BS antenna increases, uplink SNR decreases. Uplink SNR nearly same for both the schemes for $N=3$ with 40 hops and 30 hops whereas, for $N=1$, TSPTPH performs better with 40 hops. It is also noticed that, as the effect of shadowing increases, UE has to raise their SNR for uplink data transmission to achieve the same spectral efficiency.
which is function of lognormal distribution in each transmission and 2% for $K=15$, with 40 hops and 2.5% for $K=8$ with 40 hops. TSPTPH shows the TSPTPH, user hops to a new pilot signal in each transmission slot. Therefore, a group of users contaminating the channel is changing, thereby changing channel coefficient performance of TSPTPH and TSPT nearly same. From the results it is concluded that TSPTPH shows better for reuse factor $N=3$, for both $K=8$ and $K=15$ with 40 hops. TSPTPH shows average improvement in uplink performance of 4% for $K=8$ and 2.5% for $K=15$ with 40 hops whereas with 30 hops improvement is 1.5% for $K=8$ and 2% for $K=15$. For $N=1$, performance of TSPTPH and TSPT nearly same. From the results it is concluded that proposed system works well in any type of radio environment.

**Conclusion**

The efficiency of multi-cell MIMO system is severely restricted by pilot contamination. In the TSPTPH, user hops to a new pilot signal in each transmission slot. Therefore, a group of users contaminating the channel is changing, thereby changing channel coefficient $\beta$ which is function of lognormal distribution $z$ and radial distance $r$ in each transmission slot. This leads to reduction in the channel estimation error and improves uplink performance. It is noticed that the performance of TSPTPH for all three path-loss model is significantly better for reuse factor $N=3$, for both $K=8$ and $K=15$ with 40 hops. TSPTPH shows average improvement in uplink performance of 4% for $K=8$ and 2.5% for $K=15$ with 40 hops whereas with 30 hops improvement is 1.5% for $K=8$ and 2% for $K=15$. For $N=1$, performance of TSPTPH and TSPT nearly same. From the results it is concluded that proposed system works well in any type of radio environment.
In addition to this, uplink SNR required for data transmission is also calculated for spectral efficiency of 1 bit/sec/Hz. Observed results show that, as the number of BS antennas increases, uplink SNR required for data transmission decreases. It is also observed that uplink SNR increases as the shadowing effect increases for both the systems. Thus, the performance of proposed TSPTPH system enhances uplink SINR, achievable data-rate and simultaneously reduces the power consumption by reducing the uplink SNR required for data transmission at UE end.

**Appendix A**

Seven cell structure is considered each with radius $R$ and each cell is having six interfering cells as shown in Fig. 1.

$$R_1 = R_2 = R_3 = R_4 = R_5 = R_6 = R$$

SINR and data-rate depend on $\beta$, which is the function of $z$ and radial distance $r$ between UE and BS. Therefore
\[ \text{SINR} = g(R_1, R_2 \cdots R_6) \]  
\[ E(\text{SINR}) = \int_0^{R_1} \cdots \int_0^{R_6} f(R_1, R_2 \cdots R_6) \, dR_1 \cdots dR_6 \]  
where, \( f(R_1, R_2 \cdots R_6) \) is density function. Interference between two cells will be zero when pilot signals are not reuse i.e.  
\[ f(R_1, R_2 \cdots R_6) = f(R_1) f(R_2) \cdots f(R_6) \]  
If pilot signals are reused in all the neighbouring cells, at any point user position \((x_1, y_1)\) with respect to cell having radius \( R_i \) with centre \((\alpha_i, \gamma_i)\) will have probability density function of  
\[ f(R_i) = f_{\gamma_i}(x_1, y_1) \]  
\[ f(R_i) = \begin{cases} \frac{1}{\pi R_i^2} \left( (x_i - \alpha_i)^2 + (y_i - \gamma_i)^2 \right) & \text{if UE position} \leq R_i^2 \\ 0 & \text{otherwise} \end{cases} \]  
Similarly  
\[ f(R_2) = f_{\gamma_2}(x_2, y_2) \]  
\[ f(R_2) = \begin{cases} \frac{1}{\pi R_2^2} \left( (x_2 - \alpha_2)^2 + (y_2 - \gamma_2)^2 \right) & \text{if UE position} \leq R_2^2 \\ 0 & \text{otherwise} \end{cases} \]  
Where limits are  
\[ \gamma_1 - \sqrt{R_1^2 - (x_1 - \alpha_1)^2} \leq y_1 \leq \gamma_1 + \sqrt{R_1^2 + (x_1 - \alpha_1)^2} \]  
\[ f_{R_i}(x_i) = \int_{\gamma_i - \sqrt{R_i^2 - (x_i - \alpha_i)^2}}^{\gamma_i + \sqrt{R_i^2 + (x_i - \alpha_i)^2}} \frac{1}{\pi R_i^2} \, dy \]  
\[ f_{R_i}(x_i) = \frac{1}{\pi R_i^2} 2\sqrt{R_i^2 - (x_i - \alpha_i)^2} \quad \text{where,} \quad \alpha_i - R_i \leq x_i \leq \alpha_i + R_i \]  
Now considering all six interfering cells users, \( E(\text{SINR}) \) is given as  
\[ E(\text{SINR}) = \int_{\alpha_i - R_i}^{\alpha_i + R_i} \cdots \int_{\alpha_i - R_i}^{\alpha_i + R_i} g(R_1, \cdots R_6) \times \left( \frac{2}{\pi} \right)^6 \times \prod_{m=1}^{6} \sqrt{R_m^2 - (x_m - \alpha_m)^2} \, dx_1 \cdots dx_6 \]
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