Cellular System Performance Analysis Under Correlated Shadow Fading

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Overview

1. Motivation
   - Why Correlated Shadow Fading

2. Cooperative Communication for a Single-Cell Model
   - Angular and Distance Correlation Model
   - How Cooperative Communication Mitigate Shadow Fading

3. Single-Cell System Performance Analysis
   - First-Order Markov Chain Model
   - Analysis of Outage Behavior

4. Multi-Cell System Performance Analysis
   - Grid Model vs. Random Model
   - Outage Probability and Outage Durations

5. Transport Layer Protocols for Next Generation Networks
   - Motivation
   - Data-Driven End-User Congestion Prediction Algorithm

6. Conclusions and Future Work

Overview of section 1

Introduction of Fading

In general, fading can be divided into two categories: large-scale fading and small-scale fading

- Large-scale fading: shadow fading caused by obstacles (trees, buildings, etc.) in the propagation path.
- Small-scale fading: caused by multipath propagation.

Next Generation Cellular Networks:

- Millimeter wave (mmWave) frequency spectrum, between 30 and 300 GHz, will most likely be used to address the bandwidth needs.
- The small wavelengths of mmWave will result in high path loss and sensitivity to obstructions which cause large-scale shadow fading.
- Shadow fading can lead to significant received power loss over a wide area.

Correlated Shadow Fading

In general, shadow fading is approximated by an independent log-normal distribution with a standard deviation derived from empirical measurements.

This model fails to capture the spatial correlations in shadow fading.

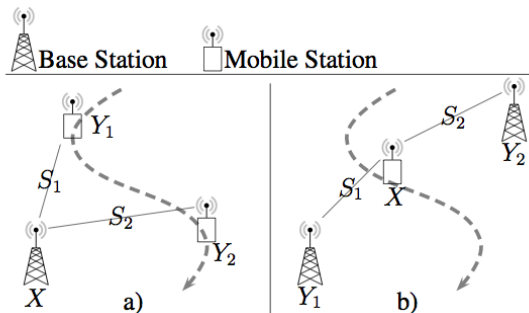A mobile user moving behind a row of tall buildings.

Shadow Fading Correlation Models

Specific Correlation Models:

- Constant Model
- Absolute Distance-Only Models: Exponential Model
- Angle-Only Models
- Separable Models: Angle-Distance Ratio and Angle-Absolute Distance

Autocorrelation and Cross-Correlation:
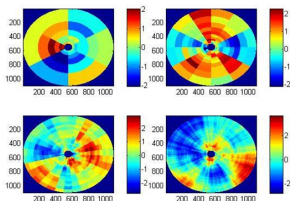
Overview of section 2

Angular and Distance Correlation Model

We consider a correlation model involves both angular and distance as follows:

$$\theta = |\angle \vec{r}_i - \angle \vec{r}_j| \in [0°, 180°], \qquad (1)$$

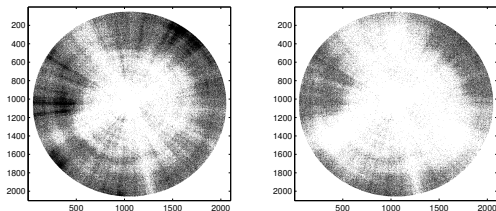$$R = |10 \log_{10} r_i/r_j| = \frac{10}{\ln 10} |\ln r_i - \ln r_j|, \qquad (2)$$

$$h(\vec{r}_i, \vec{r}_j) = max\{1 - \theta/\theta_0, 0\} \cdot max\{1 - R/R_0, 0\}. \qquad (3)$$
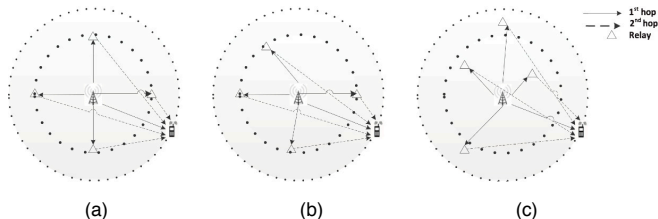

Correlated Shadowing Fields for Increasing Resolutions

Correlated Outage Fields

Correlated Outage Fields (Dark areas are outage areas while white areas are non-outage areas)

System Model

A single cell cellular deployment.

System Model and Three Different Relay Placements

- Relays are placed uniformly on a circle near the cell edge.
- Relays are randomly spaced on a circle near the cell edge.
- Relays are placed randomly in the cell.

Cooperative Communication Scheme

- First Hop: BS broadcasts signals to both relays and MS. If MS receives and decodes the signals transmitted by BS successfully, there is no need for relays to repeat the transmission. If not, relays which successfully decode the signals will participate in the second hop retransmission. Relays which cannot decode the received signal successfully, will not participate in the second hop.

- Second Hop: Among all relays which participate in the second hop, the one that has the best channel gain between the relay and MS will be chosen (this can be done by centralized control or using information from pilot signals). This relay will encode the signals again and send the coded bits to the MS.

Outage Probability of the System

The outage probability of the system:

$$P_{out_S} = P_{out_0} * (P_{out_1} + (1 - P_{out_1}) * P_{out_2}). \tag{4}$$

Simulation and Performance Evaluation
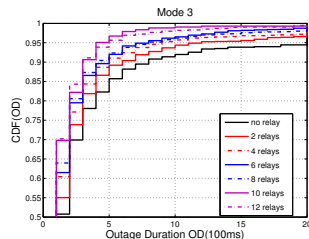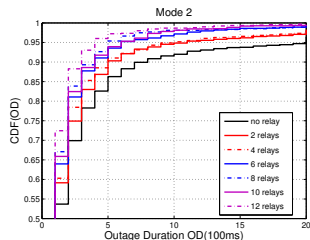
Simulation Configuration Parameters

Okuruma-Hata Model	BS Height: $100m$ Relay Height: $10m$ MS Height: $1.5m$
Correlated Shadow Fading	$R_0 : 6$ $\theta_0 : \pi/3$ $d_{min} : 50m$
Relay Placements	Three Modes with Density: 2,4,6,8,10,12
Cell Size	$R : 1000m$
MS Moving Speed	$v : 10m/s$
Radio Frequency	$f : 900MHz$
BS Transmission Power	$P : 26dbm$
SNR Requirement	$8dB$

Simulation Results and Analysis
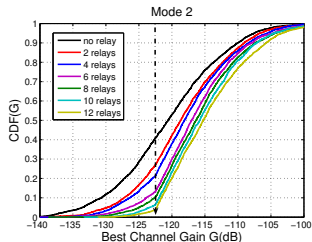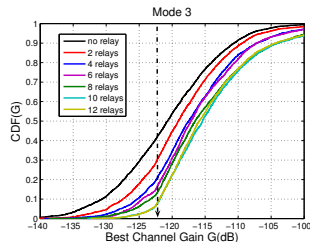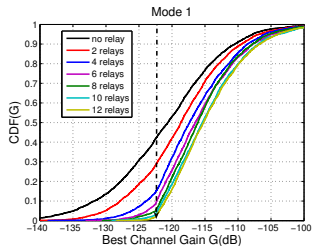
Best Channel Condition between MS and BS or Relays (corresponding to three different relay deployment modes and several different relay densities. Dashed arrow demonstrates the channel condition that satisfies the SNR requirement.
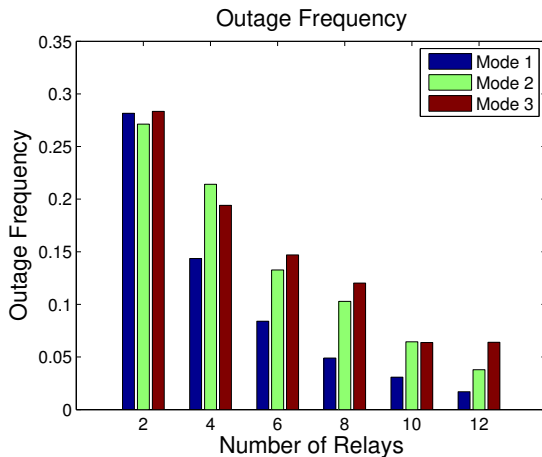
Outage Duration

Cumulative Distribution Function of Outage Duration (with Rayleigh fading)

Outage Frequency

$$f_{outage} = \frac{\text{Total Number of Outage Points}}{\text{Total Number of Simulation Points}}. \qquad (5)$$

Overview of section 3

Exponential Correlation Model

Exponential correlation model is an analytical model proposed by Gudmundson based on empirical measurements of $900MHz$ frequency.

$$\rho = e^{-\frac{d}{d_0}} \tag{6}$$

- Derived from empirical measurements.
- Does not violated physical rules.
- Correlation matrix is positive semi-dedifinite.
- Can be modeled as a first-order autoregressive process spatially.
- Has low simulation complexity.

Disadvantage

The correlation is only decided by d, θ is ignored.

Exponential Correlation

A and B are two neighboring points. Assume the shadow fading (in dB) is $N(0, \sigma^2)$.

The spatial correlation between S_A and S_B is:

$$\rho_{A,B} = \frac{E[S_A S_B]}{\sigma^2} = e^{-\frac{d_{A,B}}{d_0}} \tag{7}$$

This can be written as:

$$S_B = \rho S_A + (1 - \rho) n_A \tag{8}$$

where n_A denotes the channel noise at A.

First-Order Markov Chain Model

Lemma

Shadow fading factors of any two points that can be connected by a trajectory have jointly gaussian distribution.

Proof.

Suppose the two points are a and b, we can assume there are n positions on this trajectory $(t_1, t_2, \ldots, t_n)$. Then we have the following:

$$
\begin{aligned}
S_b &= \rho S_{t_n} + (1 - \rho)e_{t_n} \\
&= \rho(\rho S_{t_{n-1}} + (1 - \rho)e_{t_{n-1}}) + (1 - \rho)e_{t_n} = \ldots \ldots \\
&= \rho^n S_{t_1} + \sum_{i=1}^{n} \rho^{n-i}(1 - \rho)e_{t_i} \\
&= \rho^{n+1} S_a + \rho(1 - \rho)e_a + \sum_{i=1}^{n} \rho^{n-i}(1 - \rho)e_{t_i}
\end{aligned}
\tag{9}
$$

Proof Cont'd

Proof.

Let $X = \alpha S_a + \beta S_b$, the following can be derived:

$$X = \alpha S_a + \beta(\rho^{n+1} S_a + \rho(1-\rho)e_a + \sum_{i=1}^{n} \rho^{n-i}(1-\rho)e_{t_i}) \qquad (10)$$

Since S_a, e_{t_i} and e_a are all independent and Gaussian random variables, we conclude that X is also Gaussian, which implies that S_a and S_b are jointly Gaussian. $\qquad\square$

Construct Markov Chain Model

- Divide the entire shadow fading range $[-\infty, +\infty]$ into finite number of intervals $\{[-\infty, S_0], [S_0, S_1], \ldots, [S_N, +\infty]\}$. Each interval represents a state of the Markov chain model.
- Derive the state transition matrix of the Markov chain model from the probability distribution of the correlated shadow fading.
- Derive the steady-state probability from the state transition matrix of the Markov model.
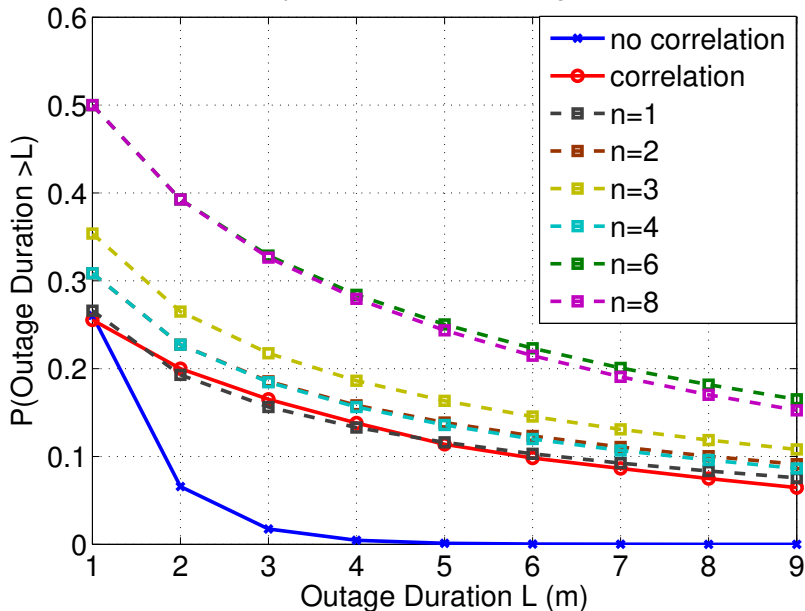
State Transition Matrix

Assume there are N states of the Markov chain model $ST_1, ST_2, \cdots, ST_N$ where ST_i corresponds to interval $(S_{i-1}, S_i]$. Then we have the state transition probability as follows:

$$
\begin{aligned}
P_{i,j} &= P(S_A \in ST_j | S_B \in ST_i) \\
&= \frac{P(S_A \in ST_j, S_B \in ST_i)}{P(S_B \in ST_i)} \\
&= \frac{\int_{S_{i-1}}^{S_i} (\int_{S_{j-1}}^{S_j} f_{(S_A | S_B = s_B)}(s_A) ds_A) f(s_B) ds_B}{\int_{S_{i-1}}^{S_i} f(s_B) ds_B}
\end{aligned}
\tag{11}
$$

Simulation Configuration Parameters

Okuruma-Hata Model	BS Height: $100m$ MS Height: $1m$
	De-Correlation Distance d_0: $20m$ Standard Deviation σ_0: $8dB$
Correlated Shadow Fading Markov Chain Model	Number of States: 50, 26, 18, 14, 10, 8 $(3 * \sigma_0/n * 2 + 2, \ n = 1, 2, 3, 4, 6, 8)$
MS Trajectory	Length: $21m$ Unit Distance δd: $1m$
Shadowing Field	$50 \times 50 m^2$
Radio Frequency	$f : 1024 MHz$
BS Transmission Power	$P : 30 dbm$
SNR Requirement	$10 dB$

Probability Distribution of Outage Duration L
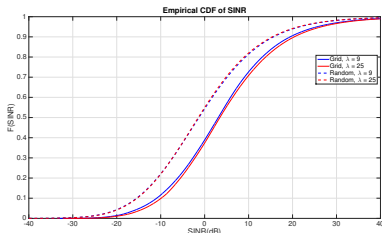
Overview of section 4

Multi-Cell System Model

- Grid Model: λ BSs are placed on a regular grid deterministically.
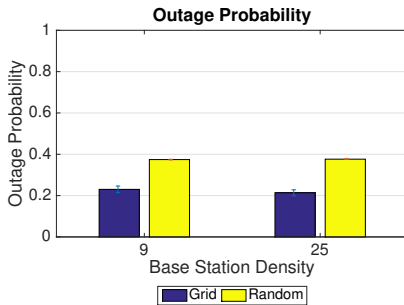- Random Model: λ BSs are placed randomly in a fixed area.

Random model and Grid model with $\lambda = 9$.

Grid Model vs. Random Model

CDF of SINR given Grid model and Random model (de-correlation distance: 20m)

Outage probability given Grid model and Random model with $\gamma = -5dB$ (de-correlation distance: 20m)

Outage Probability Simulation

Simulation Configuration Parameters
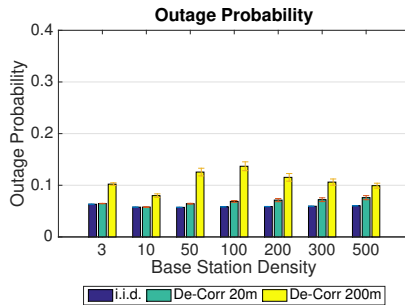
| Target Area	$1000m \times 1000m$
BS Densities	$3, 10, 50, 100, 200, 300, 500$
Path Loss Exponent	4
BS Transmission Power	$P : 40dbm$
SINR Requirement	$-5dB$
De-Correlation Distance	$20m, 200m$

Outage Probability Analysis
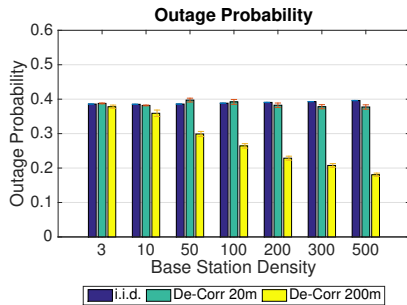
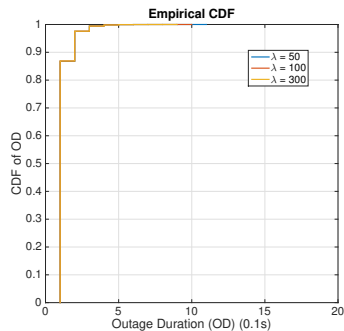Outage probability given $-5dB$ SINR threshold (MS connecting to the Nearest BS)

Outage probability given $-5dB$ SINR threshold (MS connecting to the Strongest BS)
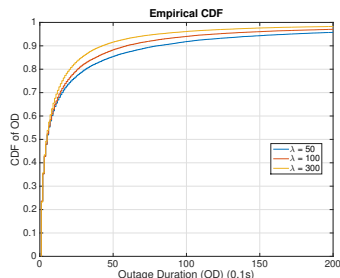
Outage Duration Analysis
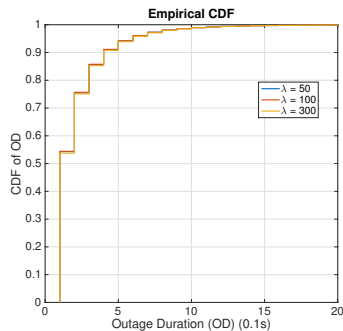
CDF of Outage Durations when the MS is connecting to the Nearest BS with independent shadow fading

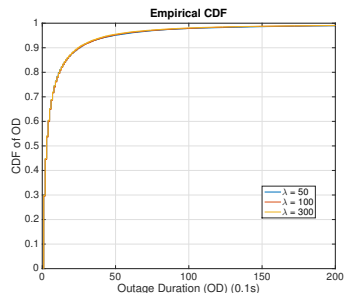CDF of Outage Durations when the MS is connecting to the Strongest BS with independent shadow shadowing

Outage Duration Analysis Cont'd

CDF of Outage Durations when MS is connecting to the Nearest BS with correlated shadowing (de-correlation distance: 200m)

CDF of Outage Durations when MS is connecting to the Strongest BS with correlated shadowing (de-correlation distance: 200m)

Overview of section 5

Motivation

MmWave Channel:

- High capacity and low delay.
- Sensitive to obstructions, which result in frequent capacity fluctuations.
- Switch between line-of-sight (LOS) state and none-line-of-sight (NLOS) state periodically.

Legacy TCP

Legacy TCP congestion control protocols might not work well with mmWave channel due to the slow congestion detection and conservative loss recovery.
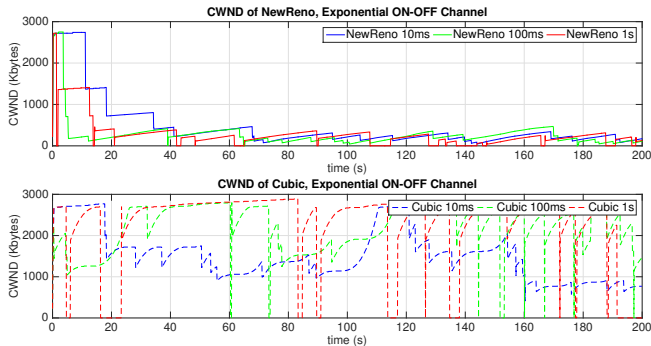
CWND dynamics given exponentially On-Off channel behavior

CWND dynamics given uniformly On-Off channel behavior

Remy

Remy is a congestion-control scheme based on prior knowledge of the network and the traffic model. It is the first computer generated congestion control protocol. A RemyCC tracks only three state variables of the end-user:

- An exponentially weighted moving average (EWMA) of the inter-arrival time between new ACKs received (ack_ewma).
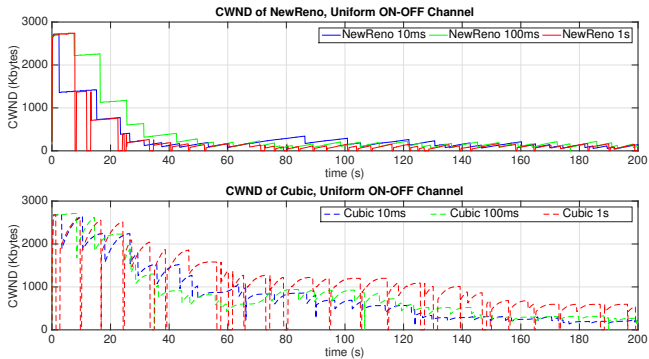- An EWMA of the time between TCP sender timestamps reflected in those ACKs (send_ewma).
- The ratio between the most recent Round Trip Time (RTT) and the minimum RTT during the current connection (rtt_ratio).

This algorithm does not require the support from any intermediate network devices. The end-user can adjust its sending speed following the pre-generated algorithm based on the above data.

Explicit Congestion Notification (ECN)

With the addition of active queue management and ECN to the network infrastructure, routers are capable of detecting congestion before queue overflows.

- An additional Congestion Experienced (CE) byte is added to the IP header of the packet to support this function.
- Routers running RED queue management can mark the CE code point with a certain probability when the average queue length satisfies a certain criteria.
- The packet with marked CE will be sent to the receiver.
- After receiving the packet, the receiver will piggyback an ACK with marked CE to notify the sender that congestion occurs in the router.

No extra traffic will be generated to overload the network. The sender will reduce the congestion window to avoid queue overflow and packet loss. Using ECN to detect congestion requires the participation of routers and receiver's feedback.

End-user Data Collection
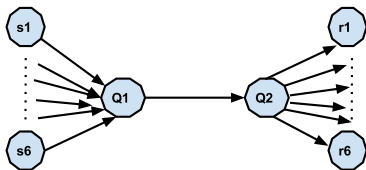
Network topology

- 6 traffic senders and 6 receivers.
- The sender s_i sends data to receiver r_i through the bottleneck link $Q1 \rightarrow Q2$.
- Both $Q1$ and $Q2$ maintain DropTail Queues.
- The queue limit is set to be the bandwidth-delay product to ensure the full utilization of the link capacity.

End-user Data Collection Cont'd

Without losing of generality, we simulated several different traffic scenarios to collect end-user data. We analyze the following scenarios:

- Scenario 1: All traffic sources are running TCP NewReno with stationary bottleneck link capacity.
- Scenario 2: All traffic sources are running TCP Cubic with stationary bottleneck link capacity.
- Scenario 3: One-half of the traffic sources is running TCP NewReno while the other half is running UDP with stationary bottleneck link capacity.
- Scenario 4: One-half of the traffic sources is running TCP Cubic while the other half is running UDP with stationary bottleneck link capacity.
- Scenario 5: Repeat scenario 1 and 2 for a periodically On-Off bottleneck link.

Feature Set

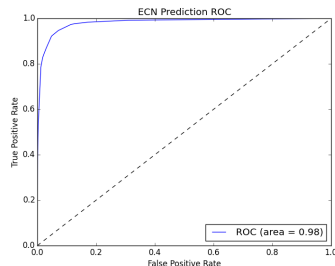Simulation Parameters

Access Link Capacity	1Gbps
BottleNeck Link Capacity	1Gbps
Access Link Delay	5ms
BottleNeck Link Delay	10ms
Queue Capacity	Bandwidth-Delay Product

- The sending time interval between two consecutive packets T_{s_i}.
- The receiving time interval between two consecutive ACKs T_{r_i}.
- An EWMA of T_{s_i} (send_ewma).
- An EWMA of T_{r_i} (ack_ewma).
- The ratio between the most recent RTT and the minimum RTT during the entire connection period (rtt_ratio).

Receiver Operating Characteristic (ROC) Curves
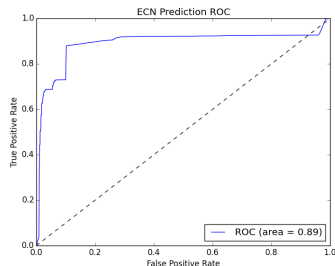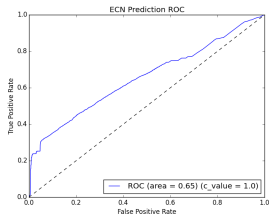

ROC curve of the Logistic Regression classifier

ROC curve of the Decision Tree classifier (max_depth = 5, min_samples_leaf = 10000)

- Logistic Regression: the decision boundary is
  $-2.32 * f_1 - 33.36 * f_2 - 12.65 * f_3 - 15.30 * f_4 + 2.03 * f_5 - 2.69 = 0$.
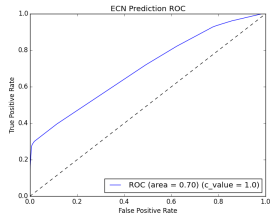- Decision Tree: the feature importance is given as [0.011703, 0.00085, 0.000000, 0.03985, 0.94760].

Receiver Operating Characteristic (ROC) Curves

Simulation Parameters

Access Link Delay	5ms, 10ms, 20ms
BottleNeck Link Delay	10ms, 20ms, 40ms
Queue Capacity	Bandwidth-Delay Product

ROC curve of Logistic Regression classifier with different propagation delays

ROC curve of Decision Tree classifier with different propagation delays

Overview of section 6

Conclusions

Conclusions:

- Cooperative communication can be used to mitigate correlated shadow fading, reduce outage frequency and probability of long-lasting outage durations.
- A first-order Markov Chain Model can be constructed to analyze the system performance under exponential correlated shadow fading.
- The performance of a multi-cell system is investigated from the outage probability and outage durations. Connecting to the BS providing highest SINR or increasing the BS density can improve the system performance.
- A data-driven end-user congestion detection algorithm is developed for stable networks.

Future Work

Future Work:

- Apply our algorithms to mitigate correlated shadow fading to the standard mmWave channel.
- Use the proposed fast end-user congestion prediction algorithm to design fast reacting TCP congestion control algorithms for stable networks.
- Extract new features and apply powerful machine learning algorithms to design fast reacting TCP congestion control algorithms for unstable networks.

Publications

Publications:

- "Shining a Light into the Darkness: How Cooperative Relay Communication Mitigates Correlated Shadow Fading", Accepted by VTC 2015 Spring.
- "How Long Before I Regain My Signal?", Accepted by CISS 2015.

Technical Report:

- "Virtual ECN: A Data-Driven End-User Congestion Prediction Algorithm".

THANKS!
QUESTIONS?