



### Quantized Conflict Graphs for Wireless Networks Optimization

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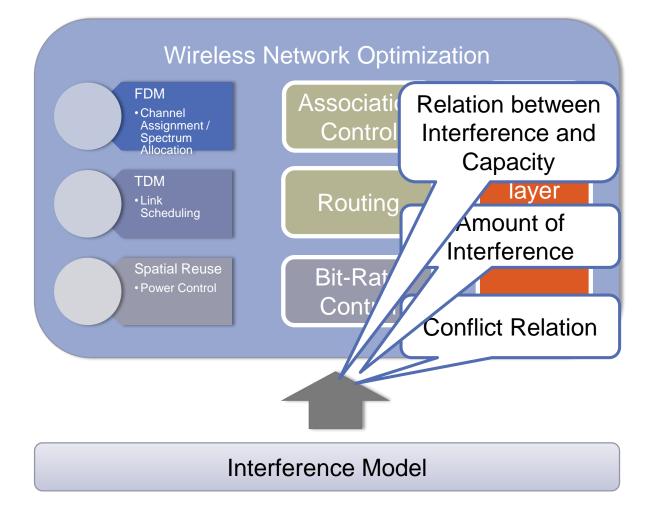
# Background

- Wireless Networks are Ubiquitous
  - WiFi
  - 3G/4G Network
  - Wireless Sensor Network
  - Wireless Mesh Network
- Wireless Medium is shared
  - The number of channels is limited
  - The bandwidth of channels is limited
  - Spatial Reuse is conflict with Coverage range



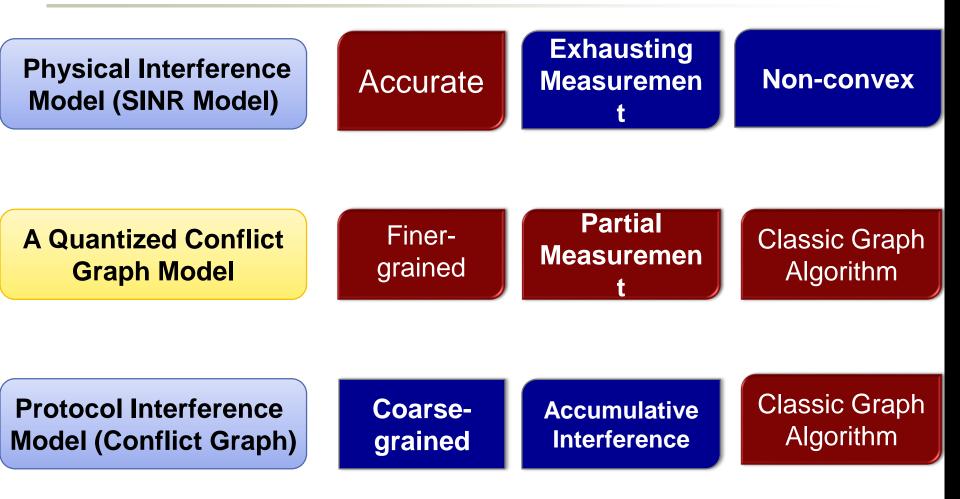


#### **Interference Model for Wireless Network Optimization**





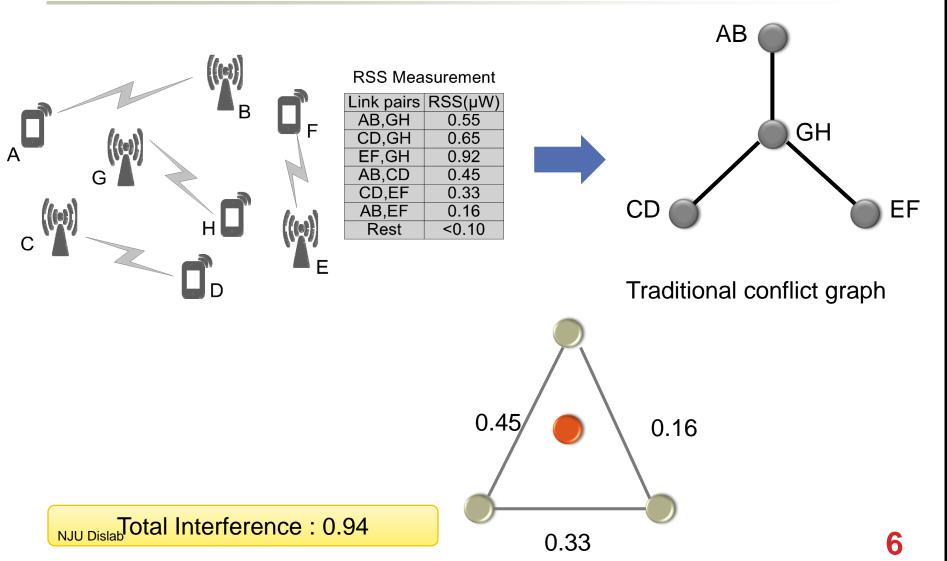
## WHY DO WE NEED A NEW MODEL







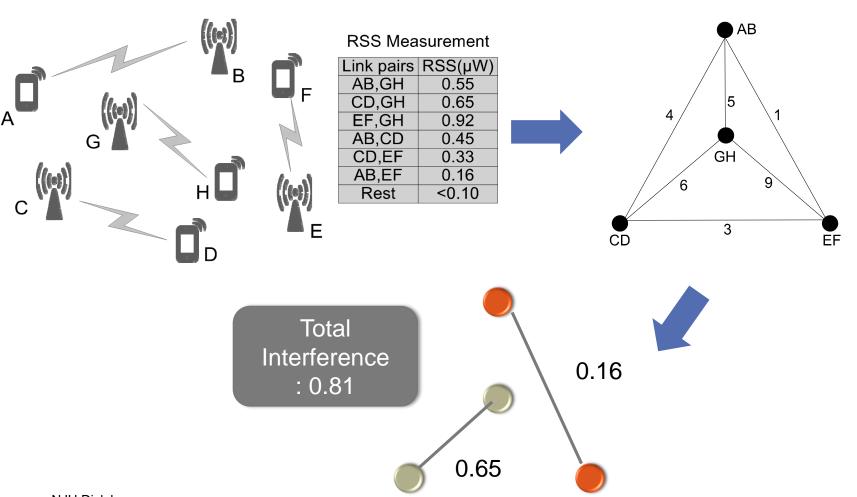
## **Motivation Example**







## **Motivation Example**





## **QCG Definition**

Quantized Conflict Graph



Pros

- Do not require an accurate RSS to get the same allocation Results
- Rough RSS representation could increase the prediction accuracy

We apply a Step function 
$$f(x) = \lfloor rac{x imes M}{C_{max}}$$

to map the RSS to the M-level weights in the graph





## **Properties of QCG**

Exhaustive signal measurements at outdoor/indoor WiFi networks

Dataset	In/out	Area (km² )	#of APs	Avg # of APs heard per location	# of measured locations
MetroFi	Outdoor	7	70	2.3	30, 991
SWIM Platform	Indoor	<1	10	5	40

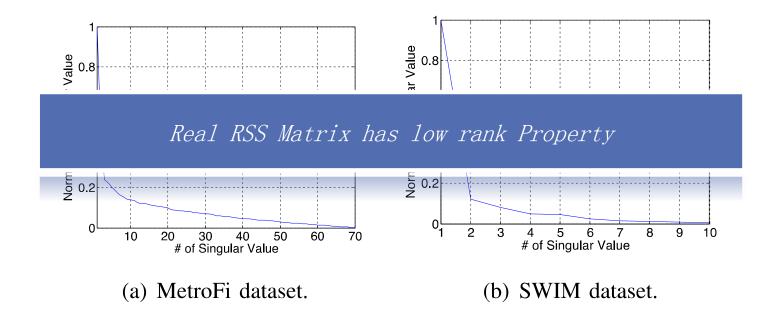
• Transformed into Quantized RSS Matrix





## **Properties of QCG**

• Explore the low-rank Property

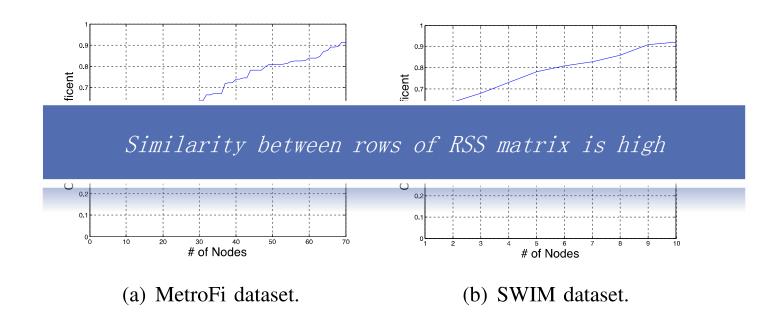






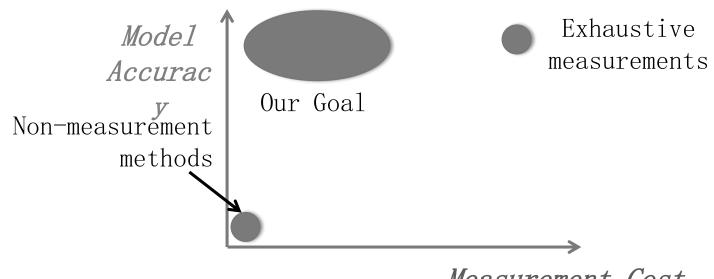
## **Properties of QCG**

• Explore the similarity between rows







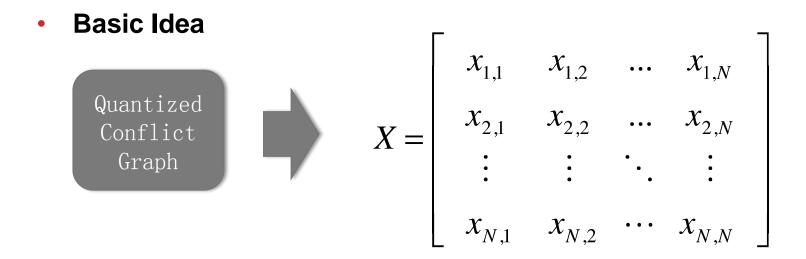


Measurement Cost

### Our approach: measure a few, predict many







Utilize the Properties of Quantized RSS Matrix

#### Treat Quantized RSS prediction as a *Matrix Completion Problem*



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## **Efficient QCG Estimation**

Low-rank Approximation  $rank(\hat{X})$  $\min$  $\hat{X}$ s.t.  $\hat{x}_{ij} = x_{ij}, \quad (i,j) \in \Omega$  $\hat{x}_{ij} \in \mathcal{C},$ Miss the similarity property between rows Rank Measurement Propagation Constraint Constraint Constraint





• Similarity

$$r_{ij} = \frac{\sum_{l=1}^{N} (X_{il} - \bar{X}_i) (X_{jl} - \bar{X}_j)}{\sqrt{\sum_{l=1}^{N} (X_{il} - \bar{X}_i)^2} \sqrt{\sum_{l=1}^{N} (X_{jl} - \bar{X}_j)^2}}$$

• We applied the idea of k-nearest neighbor here

$$\hat{x}_{ij} = \frac{\sum_{k \in N(i,j)} (S(k,j)x_{kj}))}{\sum_{k \in N(i,j)} S(k,j)}.$$

Good to deal with the Matrix with small portion of Missing Elements

CIENNEULS





- Comprehensive
  - Trivial to combine the similarity and low-rank approximation together and improve the accuracy

Use the low-rank approximation to compute a matrix X

Compute the Similarity between rows in X

Apply the linear regression for k-largest similar rows to compute weights

Compute un-known entries with the weights computed in last step





### **QCG-BASED OPTIMIZATION**

#### Interference Minimization

#### Definition

- K channels, N nodes(links)
- Minimize the total network interference



#### Scheduling

#### Definition

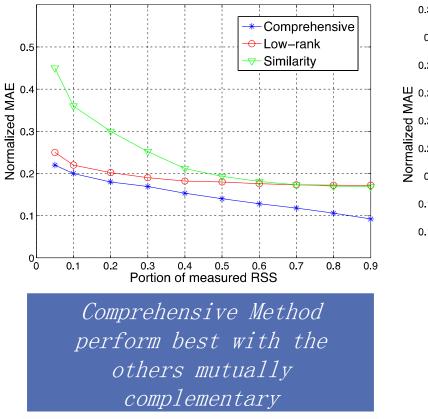
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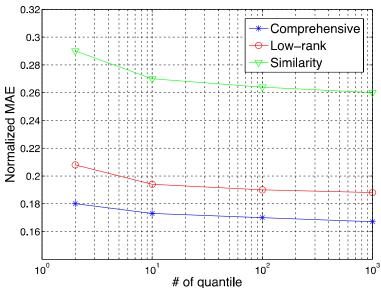
Max Weighted Independent Set





• Prediction Accuracy





Increase the number of levels larger than 10 only slightly improve the accuracy

NJU Dislab

complementar

accuracy

19





### **Evaluation**

• Wireless Network Optimization

