# **Private Release of Graph Statistics using Ladder Functions**

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## **1. Private Release of Graph Statistics**

#### **D** Private Data Release



### **3. Ladder Functions**

□ Key Idea: Change Slope Gradually

 $\int \ln(\Pr[O])$ 

 $\ln(\Pr[O])$ 

**Red** slope 1 is  $\varepsilon/I_1(g')$ 



Blue slope 0 is  $\varepsilon/LS(g) \rightarrow \varepsilon/I_0(g)$ 

 $f(g) \quad f(g')$ 

#### Differential Privacy on Graph

Differentially private algorithm injects noise into the query answer, in order to cover the maximum impact of a relationship (an edge).

query1: are nodes *a* and *b* connected?

query2: how many edges?

query3: how many triangles?



#### 2. Global and Local Sensitivity

□ Formal Definition of Differential Privacy



# $\square Summary$ $\int_{\ln(\Pr[O])} f(g)$ GS GS Iadder functions

Blue slope 1 is no larger than any  $\varepsilon/I_0(g')$ ; so it is  $\varepsilon/\max I_0(g') = \varepsilon/I_1(g)$ 

Red slope 0 is  $\varepsilon/LS(g') \rightarrow \varepsilon/I_0(g')$