# Parameter estimation with correlated photon pairs

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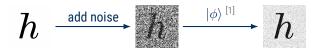


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Summary

### Motivation



**Objective:** Can coincidence measurements provide advantages in terms of precision in noisy regimes?

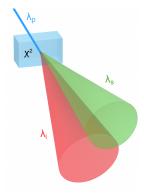


<sup>[1]</sup> Brida, Genovese, and Ruo Berchera, "Experimental Realization of Sub-Shot-Noise Quantum Imaging"



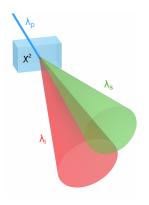


## **SPDC**

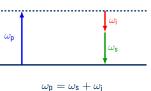








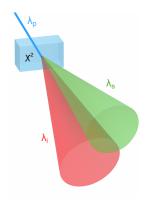
#### Energy conservation



$$\omega p = \omega_s + \omega_l$$

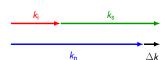


## **SPDC**



#### Energy conservation





$$\vec{k}_{\text{p}} = \vec{k}_{\text{s}} + \vec{k}_{\text{i}} - \Delta \vec{k}$$



Setup: Transmission setup



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**Setup:** Transmission setup **Parameter:** Transmittance *T* 

Motivation



**Setup:** Transmission setup **Parameter:** Transmittance T

 $\textbf{Estimator:} \ Var(\ \mathcal{T})$ 



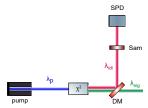
Summary

### Parameter estimation

Setup: Transmission setup **Parameter:** Transmittance T

**Estimator:** Var(T)

#### Conventional approach:



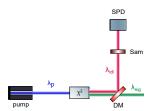




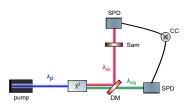
**Setup:** Transmission setup **Parameter:** Transmittance T

**Estimator:** Var(T)

#### Conventional approach:



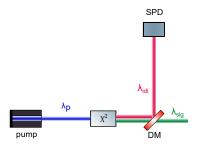
#### Coincidence approach:







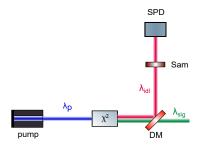
## Conventional approach



$$N_{
m tot}^{
m ref} = \eta_{
m idl} N_{
m g} + N_{
m noise}^{
m ref}$$



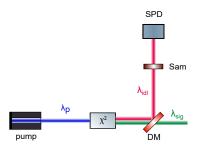
## Conventional approach



$$egin{aligned} \mathcal{N}_{\mathsf{tot}}^{\mathsf{ref}} &= \eta_{\mathsf{idI}} \, \mathcal{N}_{\mathsf{g}} + \mathcal{N}_{\mathsf{noise}}^{\mathsf{ref}} \ \mathcal{N}_{\mathsf{tot}}^{\mathsf{sam}} &= \mathcal{T} \, \eta_{\mathsf{idI}} \, \mathcal{N}_{\mathsf{g}} + \mathcal{N}_{\mathsf{noise}}^{\mathsf{sam}} \end{aligned}$$



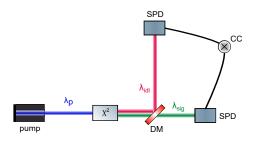
## Conventional approach



$$N_{
m tot}^{
m ref} = \eta_{
m idl} N_{
m g} + N_{
m noise}^{
m ref}$$
 $N_{
m tot}^{
m sam} = T \, \eta_{
m idl} N_{
m g} + N_{
m noise}^{
m sam}$ 
 $\Rightarrow T = rac{N_{
m tot}^{
m sam} - N_{
m noise}^{
m sam}}{N_{
m ref}^{
m ref} - N_{
m ref}^{
m ref}}$ 



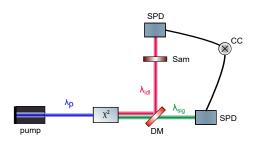
## Coincidence approach



$$N_{
m cc,tot}^{
m ref} = \eta_{
m idl} \, \eta_{
m sig} \, N_{
m g} + N_{
m ac}^{
m ref}$$

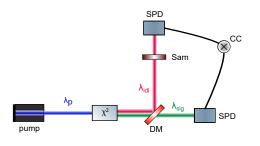


## Coincidence approach





## Coincidence approach



$$egin{align*} N_{ ext{cc,tot}}^{ ext{ref}} &= \eta_{ ext{idl}} \, \eta_{ ext{sig}} \, N_{ ext{g}} + N_{ ext{ac}}^{ ext{ref}} \ N_{ ext{cc,tot}}^{ ext{sam}} &= T \, \eta_{ ext{idl}} \, \eta_{ ext{sig}} \, N_{ ext{g}} + N_{ ext{ac}}^{ ext{sam}} \ &\Rightarrow T = rac{N_{ ext{tot,cc}}^{ ext{sam}} - N_{ ext{ac}}^{ ext{sam}}}{N_{ ext{tot,cc}}^{ ext{ref}} - N_{ ext{ac}}^{ ext{ref}}} \end{aligned}$$



### Transmittance model

### Conventional approach:

$$\mathsf{Var}(\mathcal{T}) = (\eta_{\mathsf{idl}} \ \mathsf{N_g})^{-2} \left[ \mathsf{Var}(\mathit{N_{\mathsf{tot}}^{\mathsf{sam}}}) + \mathsf{Var}(\mathit{N_{\mathsf{noise}}^{\mathsf{sam}}}) + \mathcal{T}^2 \Big[ \mathsf{Var}(\mathit{N_{\mathsf{tot}}^{\mathsf{ref}}}) + \mathsf{Var}(\mathit{N_{\mathsf{noise}}^{\mathsf{ref}}}) \Big] \right]$$

### Coincidence approach:

$$\mathsf{Var}(\mathcal{T}) = \left(\eta_{\mathsf{sig}} \, \eta_{\mathsf{idl}} \, \mathsf{N}_{\mathsf{g}}\right)^{-2} \left[ \mathsf{Var}\big(\mathsf{N}_{\mathsf{tot},\mathsf{cc}}^{\mathsf{sam}}\big) + \mathsf{Var}(\mathsf{N}_{\mathsf{ac}}^{\mathsf{sam}}) + \, \mathcal{T}^2 \Big[ \mathsf{Var}\big(\mathsf{N}_{\mathsf{tot},\mathsf{cc}}^{\mathsf{ref}}\big) + \mathsf{Var}\big(\mathsf{N}_{\mathsf{ac}}^{\mathsf{ref}}\big) \Big] \right]$$

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## Photon statistics







Summary

### Photon statistics

Poisson distribution (coherent light): [2][3]

$$\mathcal{P}(n) = \frac{\langle n \rangle^n}{n!} e^{-\langle n \rangle}$$
 $Var(n) = \langle n \rangle$ 

Results



### Photon statistics

Poisson distribution (coherent light): [2][3]

$$\mathcal{P}(n) = \frac{\langle n \rangle^n}{n!} e^{-\langle n \rangle}$$
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<sup>[3]</sup> Fouche, "Detection and False-Alarm Probabilities for Laser Radars That Use Geiger-mode Detectors"





<sup>[2]</sup> Kim et al., "Photon-Counting Statistics-Based Support Vector Machine with Multi-Mode Photon Illumination for Quantum Imaging"

Results

Summary

## Photon statistics

Poisson distribution (coherent light): [2][3]

$$\mathcal{P}(n) = \frac{\langle n \rangle^n}{n!} e^{-\langle n \rangle}$$

$$Var(n) = \langle n \rangle$$

multi-mode Bose-Einstein distribution (thermal light): [2]

$$\mathcal{P}_{m}(n) = \frac{(n+m-1)!}{(m-1)! \, n!} \frac{m^{m} \langle n \rangle^{n}}{(m+\langle n \rangle)^{n+m}}$$

$$\operatorname{Var}(n) = \langle n \rangle \left( 1 + \frac{\langle n \rangle}{m} \right)$$

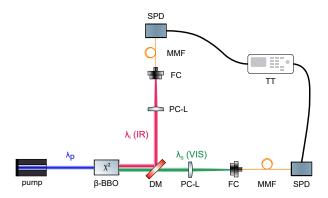


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Motivation Theory Experiment Results Simulation Summary

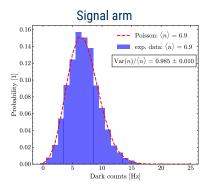
## Experimental setup







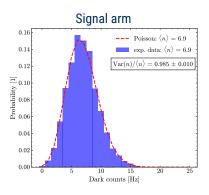


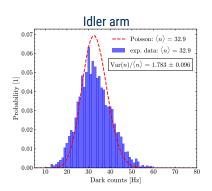




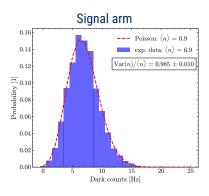


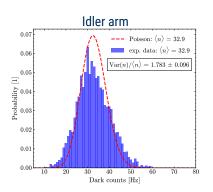












$$Var(N_{noise}) = 1.8 \cdot \langle N_{noise} \rangle$$





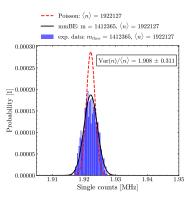
## Single counts



Motivation Theory Experiment Results Simulation Summary

## Single counts

#### Signal arm

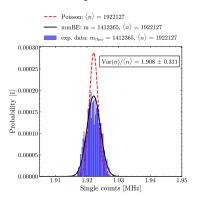




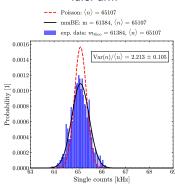


## Single counts

#### Signal arm



#### Idler arm

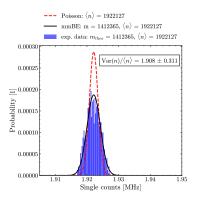




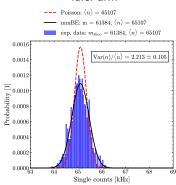


## Single counts

#### Signal arm



#### Idler arm

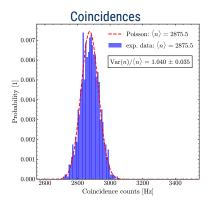


$$Var(N_{noise}) = 2.2 \cdot \langle N_{noise} \rangle$$



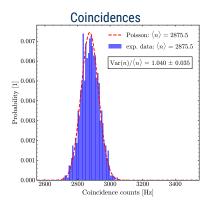








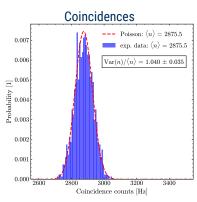




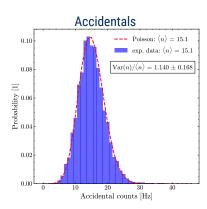
$$\text{Var}(\textit{N}_{\text{cc}}) = \langle \textit{N}_{\text{cc}} \rangle$$



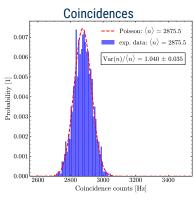




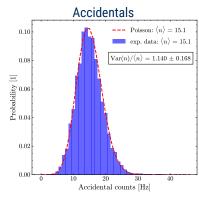
$$Var(N_{cc}) = \langle N_{cc} \rangle$$







$$Var(N_{cc}) = \langle N_{cc} \rangle$$



$$Var(N_{ac}) = \langle N_{ac} \rangle$$





Summary

### Simulation

### Conventional approach:

$$\mathsf{Var}(\mathcal{T}) = (\eta_{\mathsf{idl}} \, \mathsf{N}_{\mathsf{g}})^{-2} \left[ \mathsf{Var}(\mathsf{N}_{\mathsf{tot}}^{\mathsf{sam}}) + \mathsf{Var}\big(\mathsf{N}_{\mathsf{noise}}^{\mathsf{sam}}\big) + \mathcal{T}^2 \Big[ \mathsf{Var}\big(\mathsf{N}_{\mathsf{tot}}^{\mathsf{ref}}\big) + \mathsf{Var}\big(\mathsf{N}_{\mathsf{noise}}^{\mathsf{ref}}\big) \Big] \right]$$



Summary

### Simulation

### Conventional approach:

$$\text{Var}(\mathcal{T}) = (\eta_{\text{idl}} \, \textit{N}_{\text{g}})^{-2} \left[ 2.2 \cdot \langle \textit{N}_{\text{tot}}^{\text{sam}} \rangle + 1.8 \cdot \langle \textit{N}_{\text{noise}}^{\text{sam}} \rangle + \mathcal{T}^2 \Big[ 2.2 \cdot \langle \textit{N}_{\text{tot}}^{\text{ref}} \rangle + 1.8 \cdot \langle \textit{N}_{\text{noise}}^{\text{ref}} \rangle \Big] \right]$$

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### Coincidence approach:

$$\mathsf{Var}(\mathcal{T}) = \left(\eta_{\mathsf{sig}} \ \eta_{\mathsf{idl}} \ \mathsf{N}_{\mathsf{g}}\right)^{-2} \left[ \mathsf{Var}\big(\mathsf{N}_{\mathsf{tot},\mathsf{cc}}^{\mathsf{sam}}\big) + \mathsf{Var}(\mathsf{N}_{\mathsf{ac}}^{\mathsf{sam}}) + \mathcal{T}^{2} \Big[ \mathsf{Var}\big(\mathsf{N}_{\mathsf{tot},\mathsf{cc}}^{\mathsf{ref}}\big) + \mathsf{Var}\big(\mathsf{N}_{\mathsf{ac}}^{\mathsf{ref}}\big) \Big] \right]$$

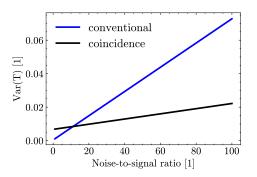
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### Coincidence approach:

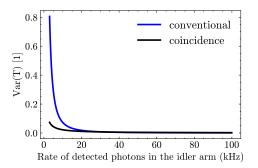
$$\mathrm{Var}(\mathit{T}) = \left(\eta_{\mathrm{sig}} \, \eta_{\mathrm{idl}} \, \mathit{N}_{\mathrm{g}}\right)^{-2} \left[ \langle \mathit{N}_{\mathrm{tot,cc}}^{\mathrm{sam}} \rangle + \langle \mathit{N}_{\mathrm{ac}}^{\mathrm{sam}} \rangle + \mathit{T}^{2} \Big[ \langle \mathit{N}_{\mathrm{tot,cc}}^{\mathrm{ref}} \rangle + \langle \mathit{N}_{\mathrm{ac}}^{\mathrm{ref}} \rangle \Big] \right]$$





Parameter	Value
$\eta_{\text{idl}}$ (%)	0.09
$\eta_{sig}$ (%)	2.6
R <sub>idl</sub> (kHz)	10
R <sub>noise,idl</sub> (kHz)	10 - 1000
R <sub>noise,sig</sub> (Hz)	7
T (1)	0.9





Parameter	Value
η <sub>idl</sub> (%)	0.09
$\eta_{sig}$ (%)	2.6
R <sub>idl</sub> (kHz)	3 - 100
R <sub>noise,idl</sub> (kHz)	1000
R <sub>noise,sig</sub> (Hz)	7
T (1)	0.9



Results

Summary

## Summary and Outlook

Can coincidence measurements provide more precise results?

#### Achievements:

- Model for variance of transmittance
- Experimental verification of photon statistics
- Found regions where coincidence approach offers advantages





## Summary and Outlook

### Git repository

#### public accessible:

https://git.tpi.uni-jena.de/mstnhsr/latexbeamer\_corporatedesign

### Feedback

marc.steinhauser@uni-jena.de





## Slide title in Palatino Linotype Font

block environment (lower-case b)

#### itemize:

- First Level
  - Second Level

Third Level has no item mark

### Block environment (upper-case B)

#### enumerate:

- First Level
  - 1.1 Second Level
    - 1.1.1 Third Level



