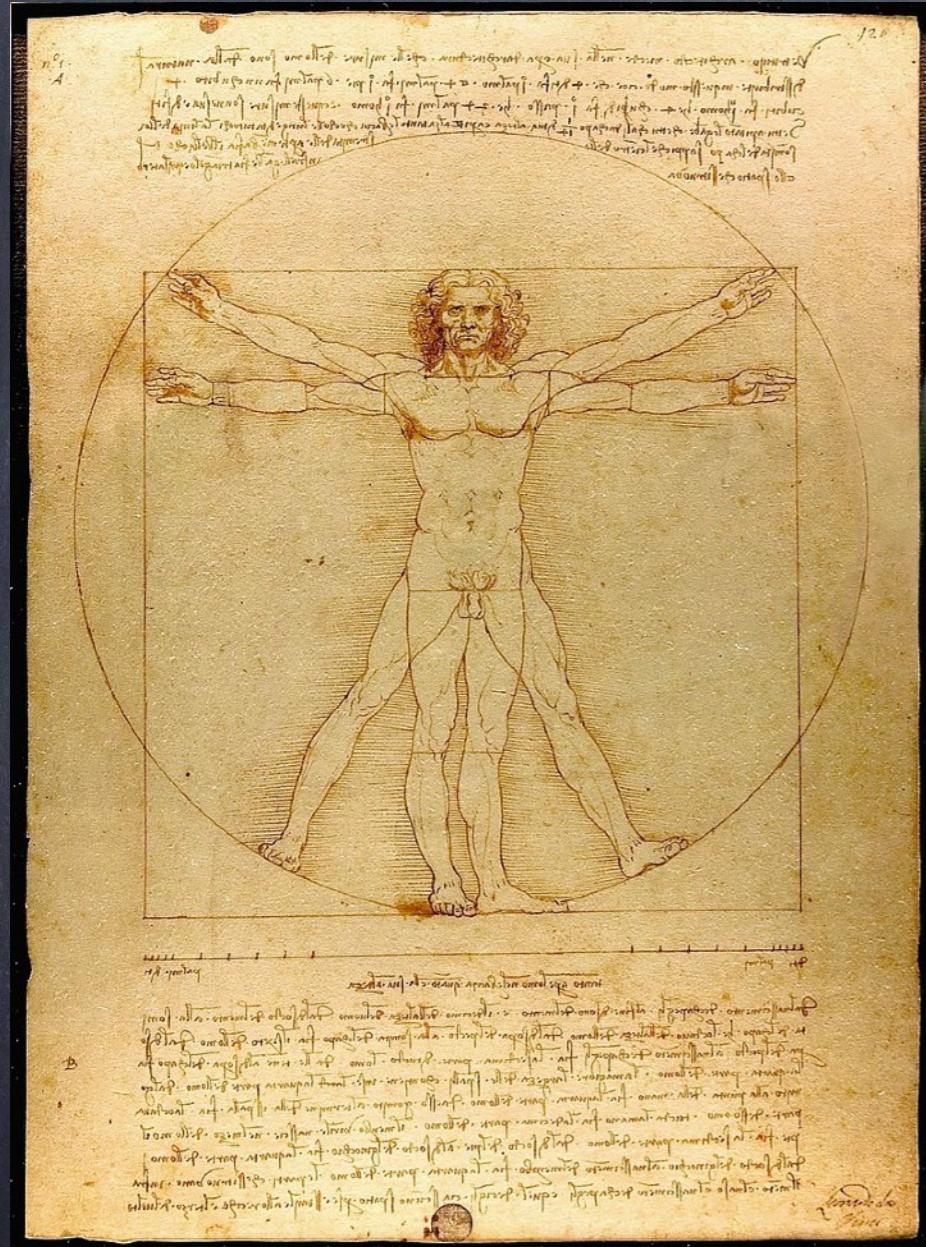




# A primer on (quantum) metrology



NEW Quantum Optics group  
Dipartimento di Scienze  
Università Roma Tre, Italy



Metrology (n): the scientific  
study of measurement.  
From Greek metron

Омне trinum perfectum

Omne trinum perfectum

probe

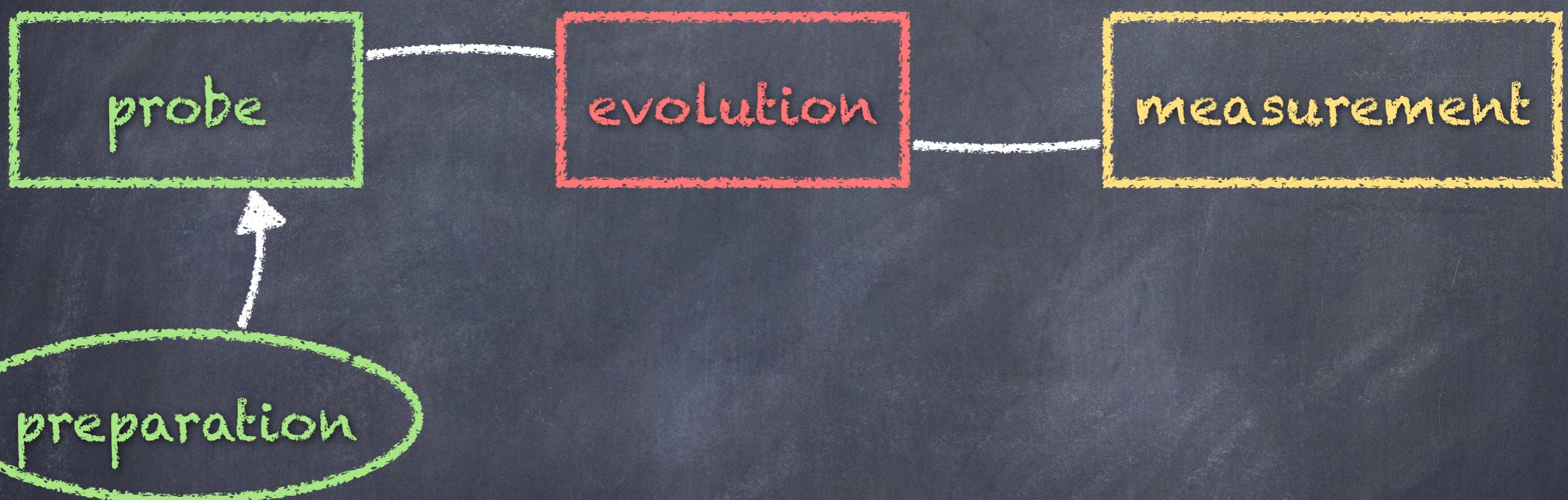
Omne trinum perfectum



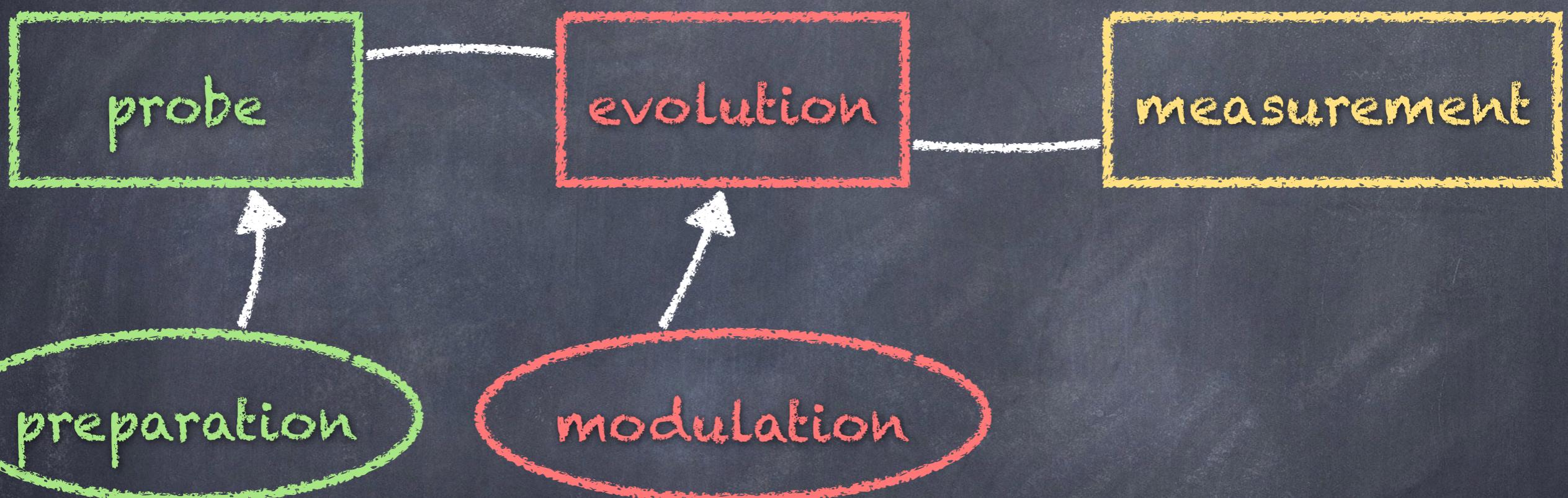
# Omne trinum perfectum



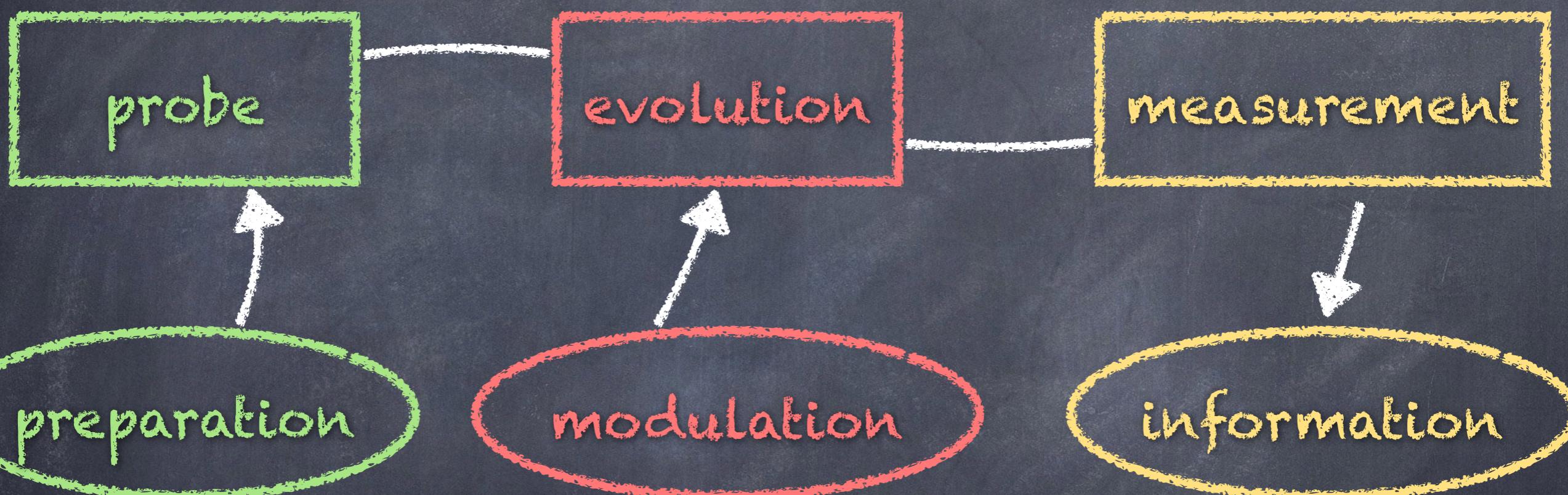
# Omne trinum perfectum



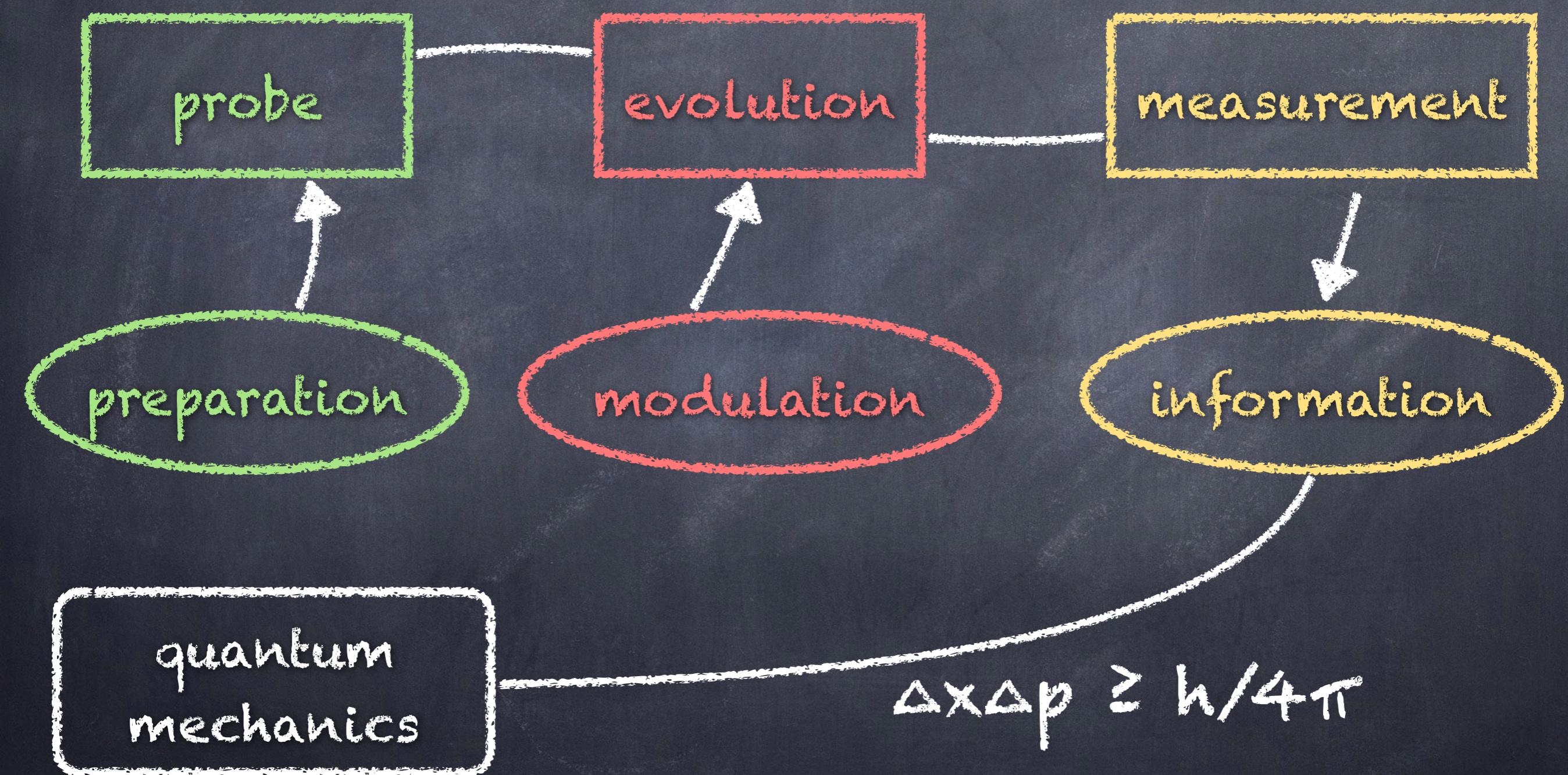
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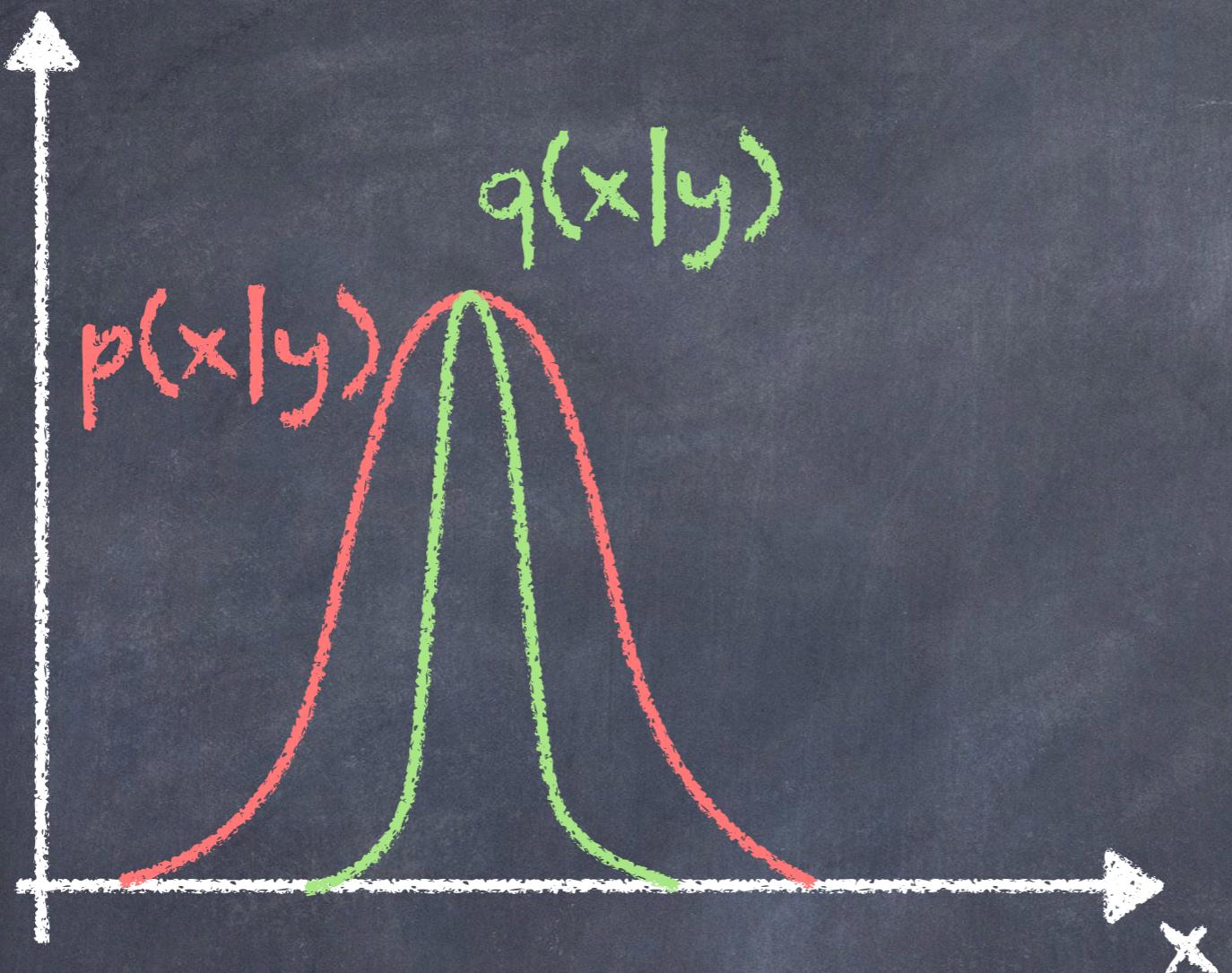










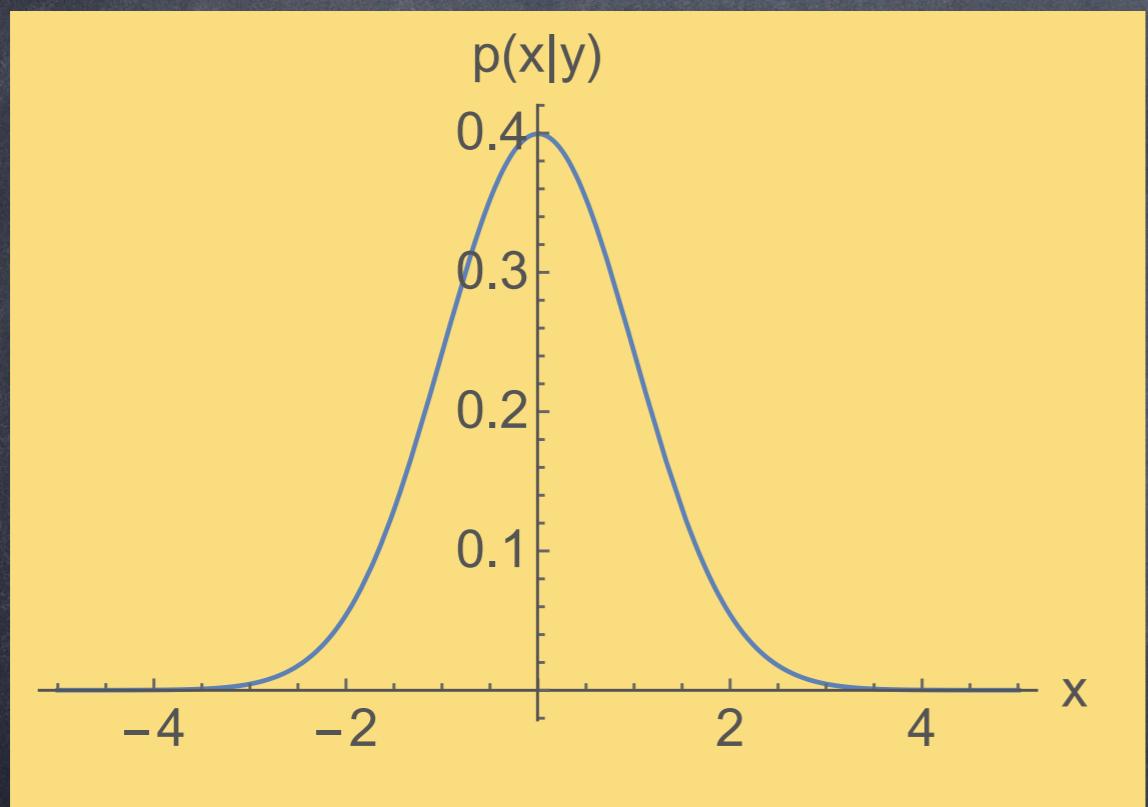




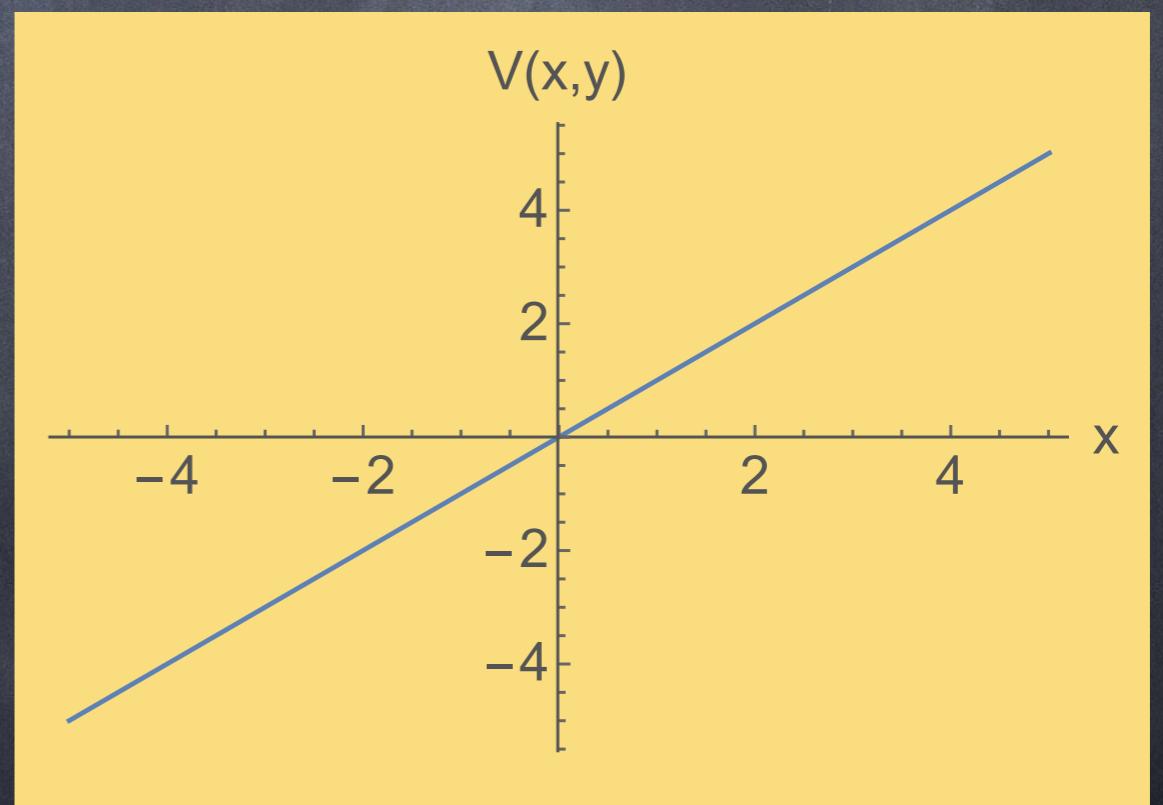
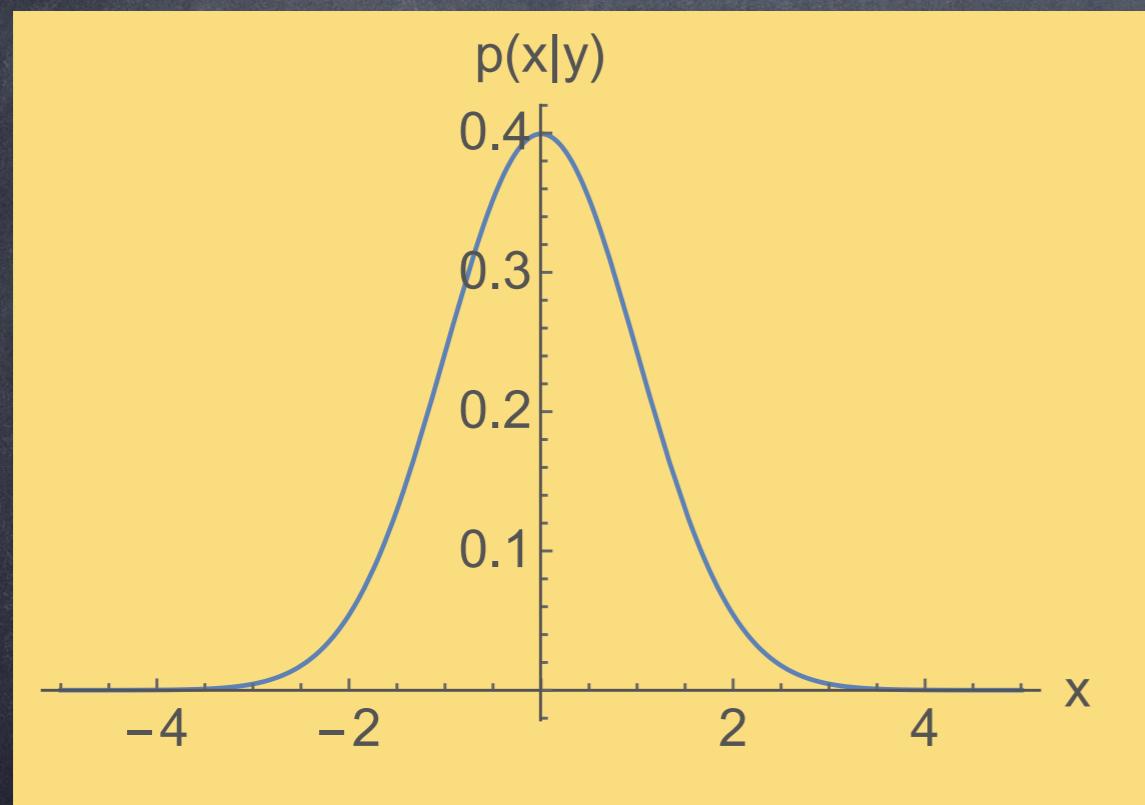
$q(x|y)$  is better than  $p(x|y)$

score function

# score function

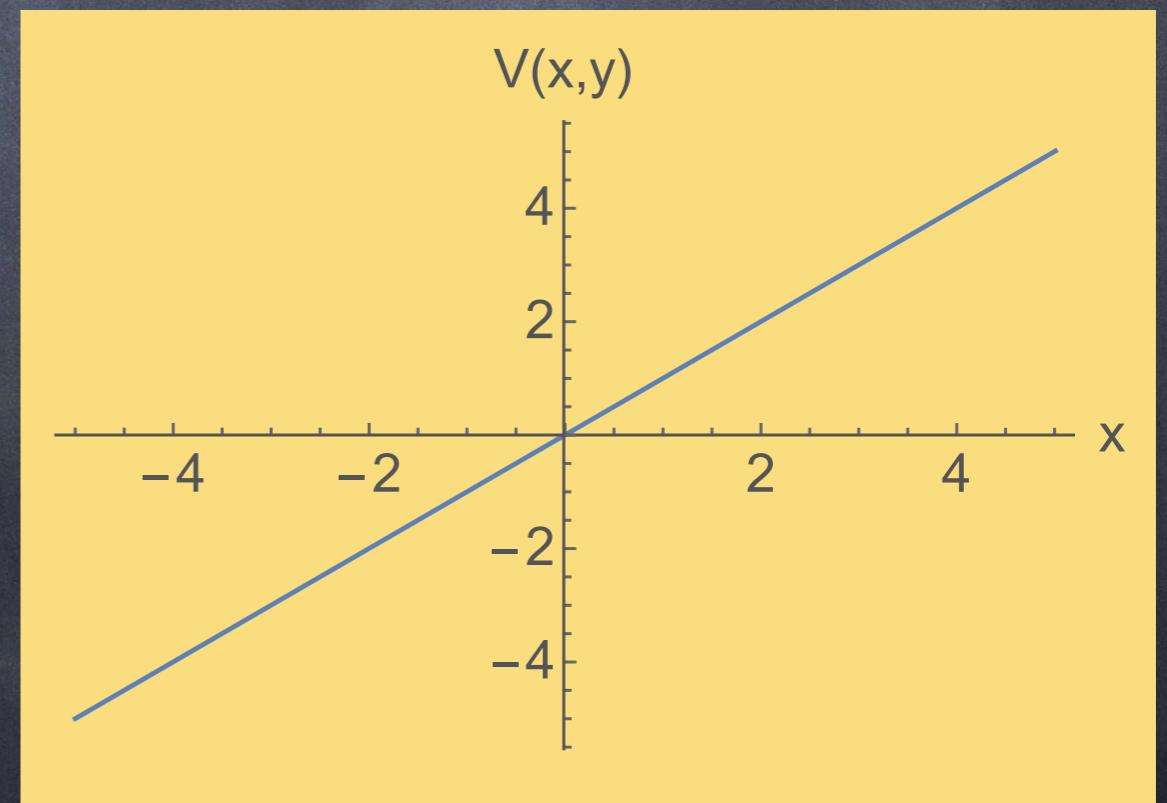
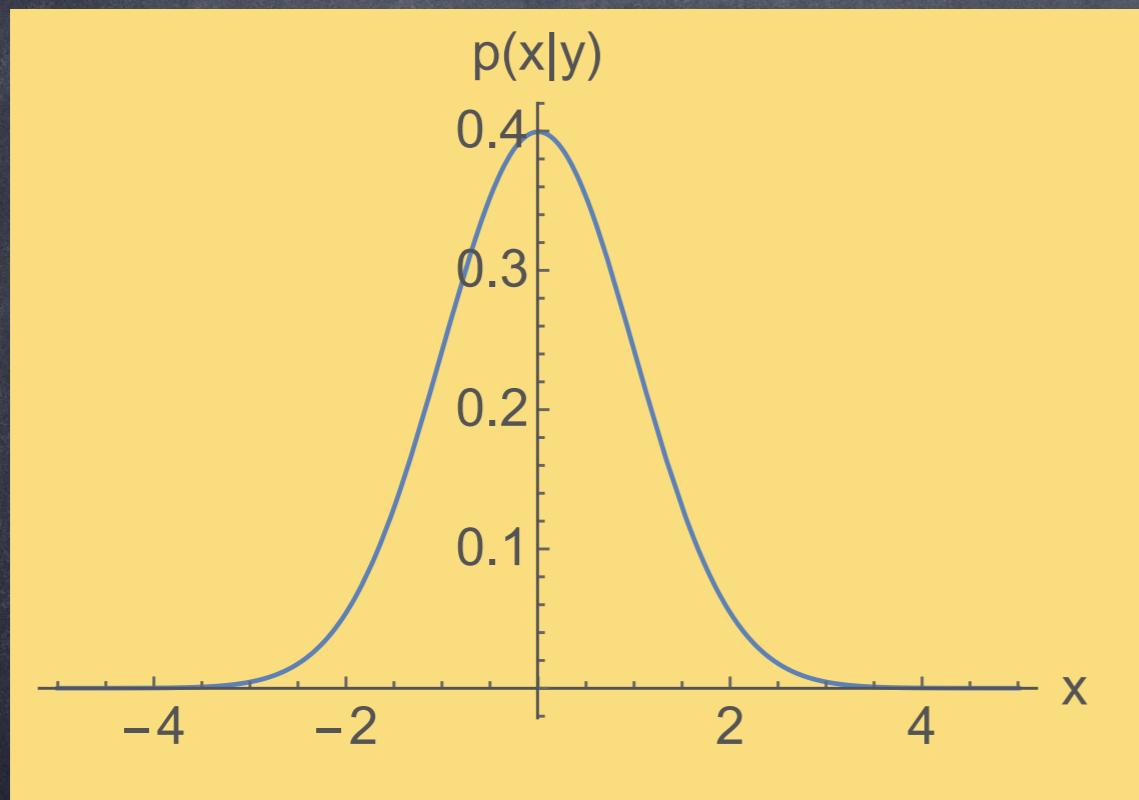


# score function



# score function

$$V(x, y) = \partial_y \log(p(x, y)) = \frac{\partial_y p(x, y)}{p(x, y)}$$



Fisher information

# Fisher information

$$\mathcal{I}(y) = \mathbf{E}[V(x, y)^2] = \int \frac{(\partial_y p(x, y))^2}{p(x, y)} dx$$

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$$\mathcal{I}(y) = \frac{1}{\sigma^2}$$

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Cramér–Rao bound

# Fisher information

$$\mathcal{I}(y) = \mathbf{E}[V(x, y)^2] = \int \frac{(\partial_y p(x, y))^2}{p(x, y)} dx$$

[lots of maths which is essentially a clever use of the Cauchy Schwartz inequality]

$$\mathcal{I}(y) \geq \frac{1}{\text{Var}[y]}$$

Cramér–Rao bound

what if we cumulate M experiments

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$$\mathcal{I}_{\text{tot}}(y) \geq M\mathcal{I}(y)$$

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Cramér-Rao bound

$$\text{Cramér-Rao bound} \quad \mathcal{I}(y) \geq \frac{1}{M\text{Var}[y]}$$

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$x_1, x_2, x_3, \dots, x_M$

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$y_{\text{est}}$



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$$\mathbf{E}[y_{\text{est}} - y] = 0$$

Cramér–Rao bound

$$\mathcal{I}(y) \geq \frac{1}{M\text{Var}[y]}$$

1. valid for unbiased estimators

$$x_1, x_2, x_3, \dots, x_M$$

$$y_{\text{est}}$$



$$\mathbf{E}[y_{\text{est}} - y] = 0$$

$$\text{Var}[y] = \mathbf{E}[(y_{\text{est}} - y)^2]$$

Cramér-Rao bound

$$\mathcal{I}(y) \geq \frac{1}{M\text{Var}[y]}$$

2. valid for local estimation

$$x_1, x_2, x_3, \dots, x_M$$

$$y_{\text{est}}$$



$$\mathbf{E}[y_{\text{est}} - y] = 0$$

$$\text{Var}[y] = \mathbf{E}[(y_{\text{est}} - y)^2]$$

Cramér–Rao bound

$$\mathcal{I}(y) \geq \frac{1}{M\text{Var}[y]}$$

3. valid asymptotically

$$x_1, x_2, x_3, \dots, x_M$$

$$y_{\text{est}}$$



$$\mathbf{E}[y_{\text{est}} - y] = 0$$

$$\text{Var}[y] = \mathbf{E}[(y_{\text{est}} - y)^2]$$

Cramér-Rao bound

$$\mathcal{I}(y) \geq \frac{1}{M\text{Var}[y]}$$

3. valid asymptotically

$x_1, x_2, x_3, \dots, x_M$



$$\mathbf{E}[y_{\text{est}} - y] = 0$$

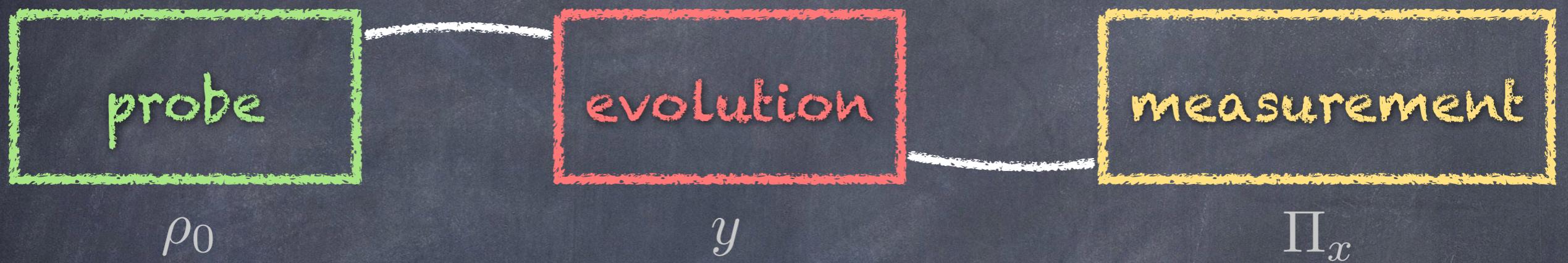
$$\text{Var}[y] = \mathbf{E}[(y_{\text{est}} - y)^2]$$

Where do we get  $p(x|y)$  from?

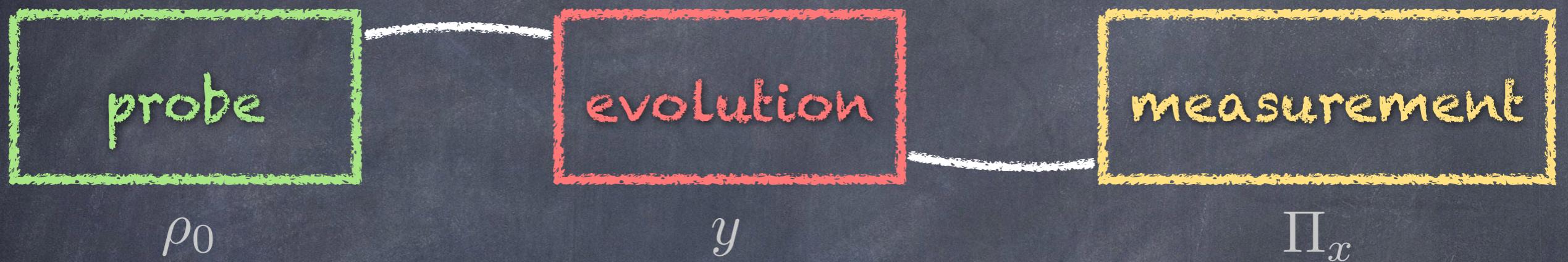
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- which POVM?
- which state?

measure in the direction in which the state curves the most

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$$2\partial_y \rho_y = L_y \rho_y + \rho_y L_y$$

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$\mathcal{I}_Q(y) = \text{Tr}[\rho_y L_y^2]$  Quantum Fisher Information

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$\mathcal{I}_Q(y) \geq \mathcal{I}(y)$  Quantum Cramér–Rao bound

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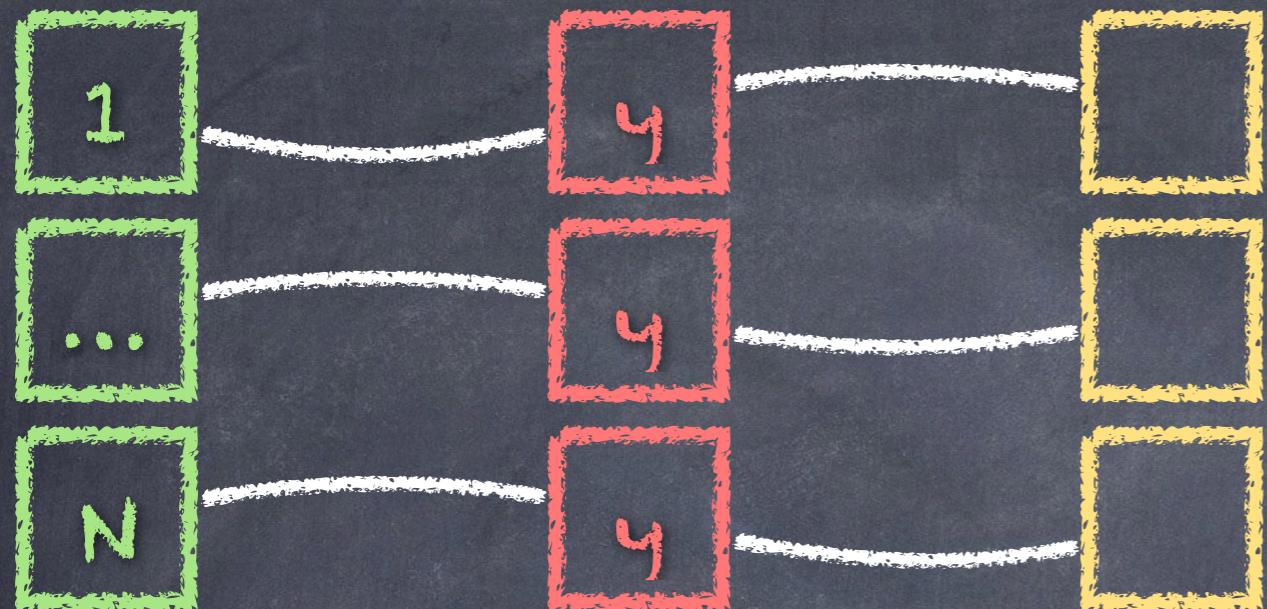
$$2\partial_y \rho_y = L_y \rho_y + \rho_y L_y$$

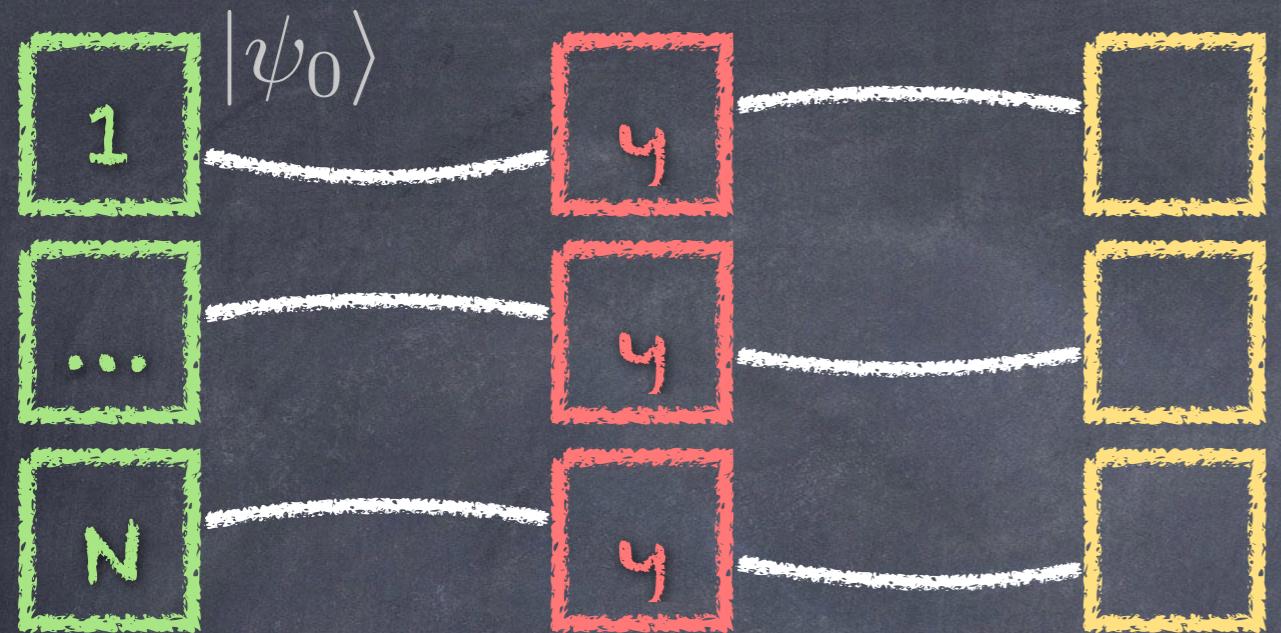
$$\mathcal{I}_Q(y) \geq \mathcal{I}(y) \quad \text{Quantum Cramér-Rao bound}$$

The optimal measurement is the SLD

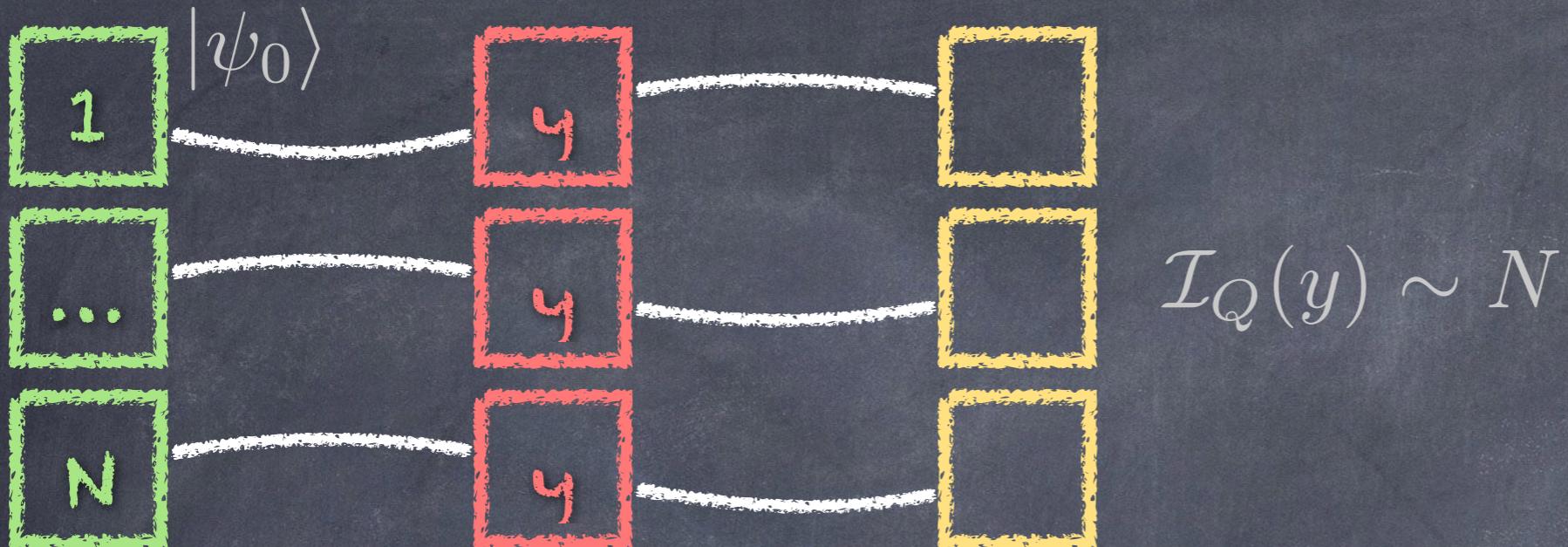
$$\mathcal{I}_Q(y) = \frac{1}{M\text{Var}[y]}$$



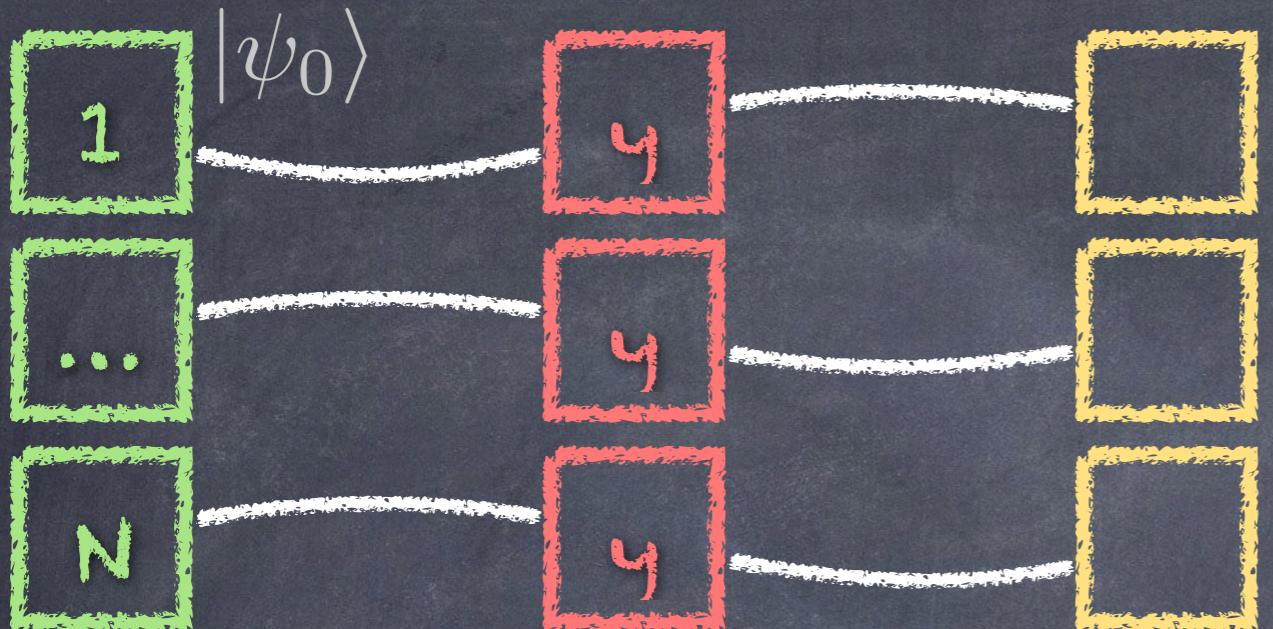




$$U_y = \exp(-iH_y)$$

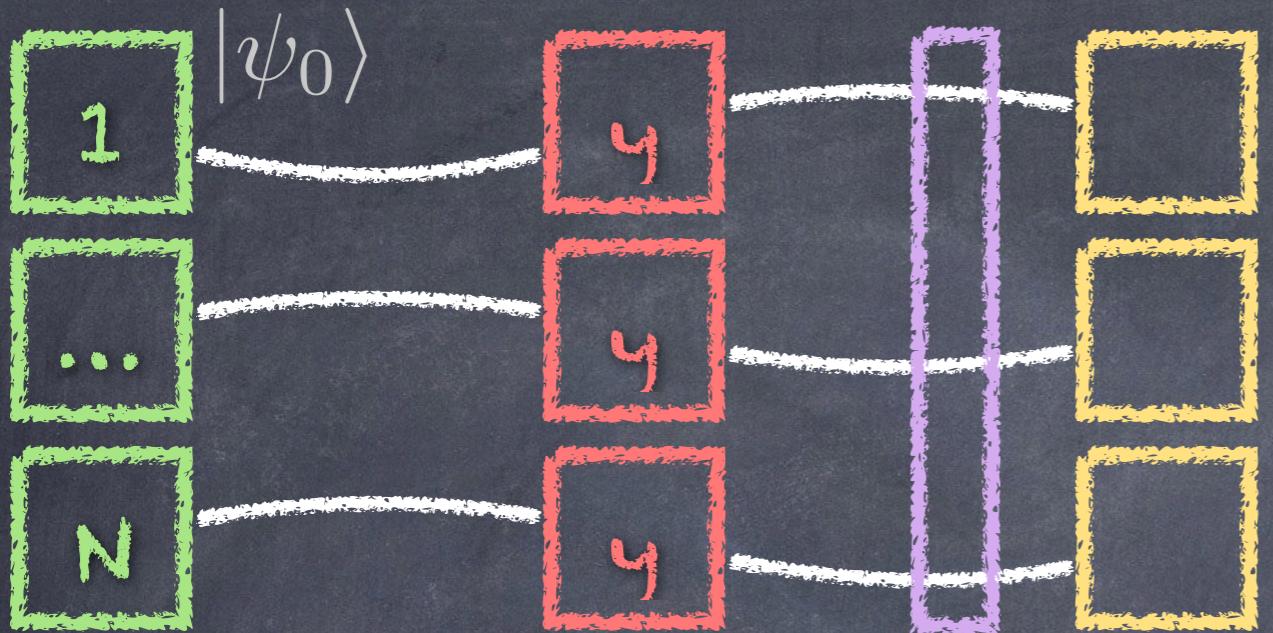


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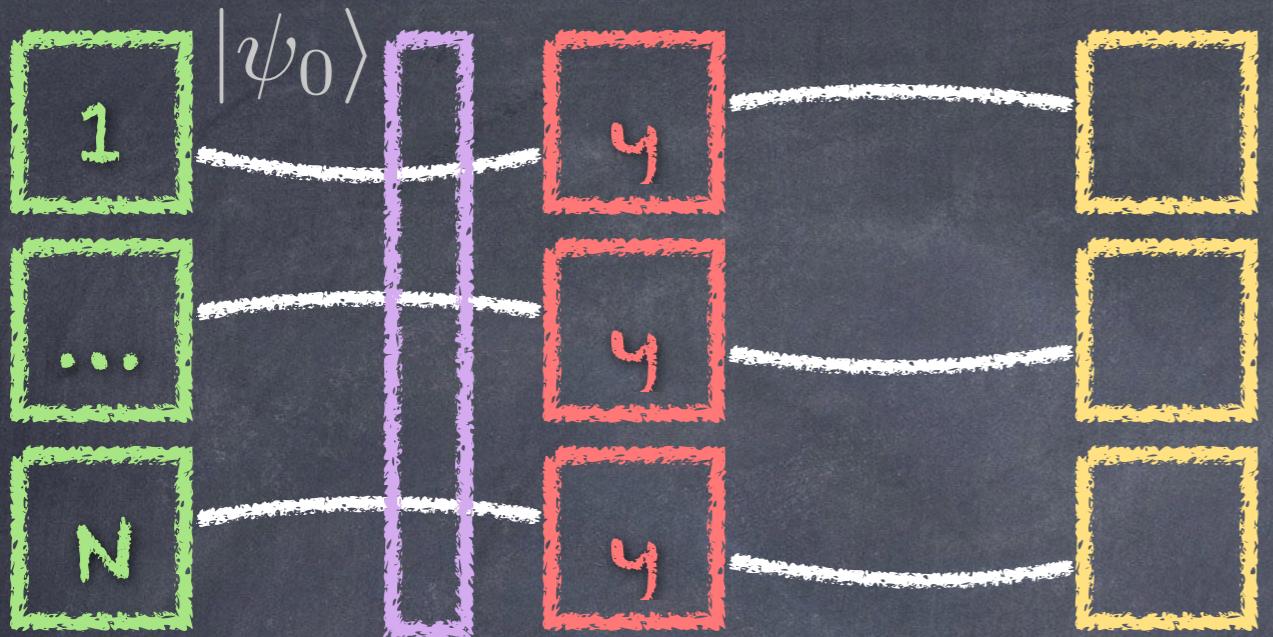
A diagram on a chalkboard showing two overlapping circles, one green and one yellow. Between the circles, the text  $I_Q(y) \sim N$  is written.



$$U_y = \exp(-iH y)$$

C       $\mathcal{I}_Q(y) \sim N$

Q       $\mathcal{I}_Q(y) \sim N$

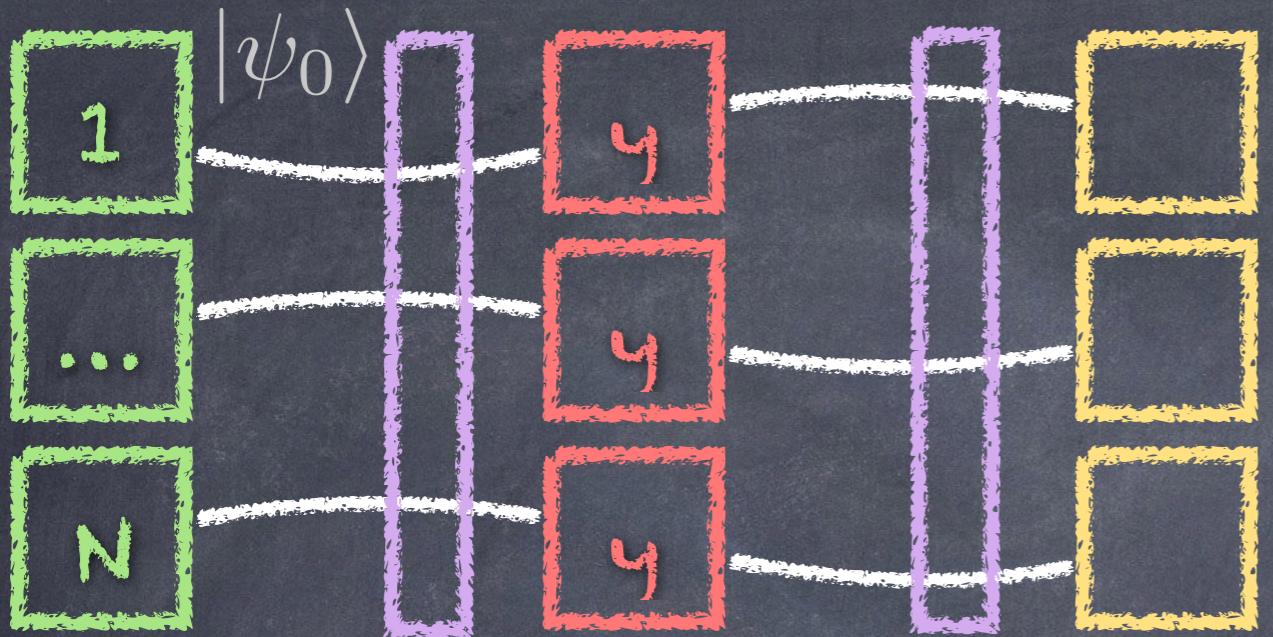


$$U_y = \exp(-iH y)$$

Two cases are shown for the distribution of  $\mathcal{I}_Q(y)$ :

- If  $C < Q$ , then  $\mathcal{I}_Q(y) \sim N$ .
- If  $C > Q$ , then  $\mathcal{I}_Q(y) \sim N^2$ .

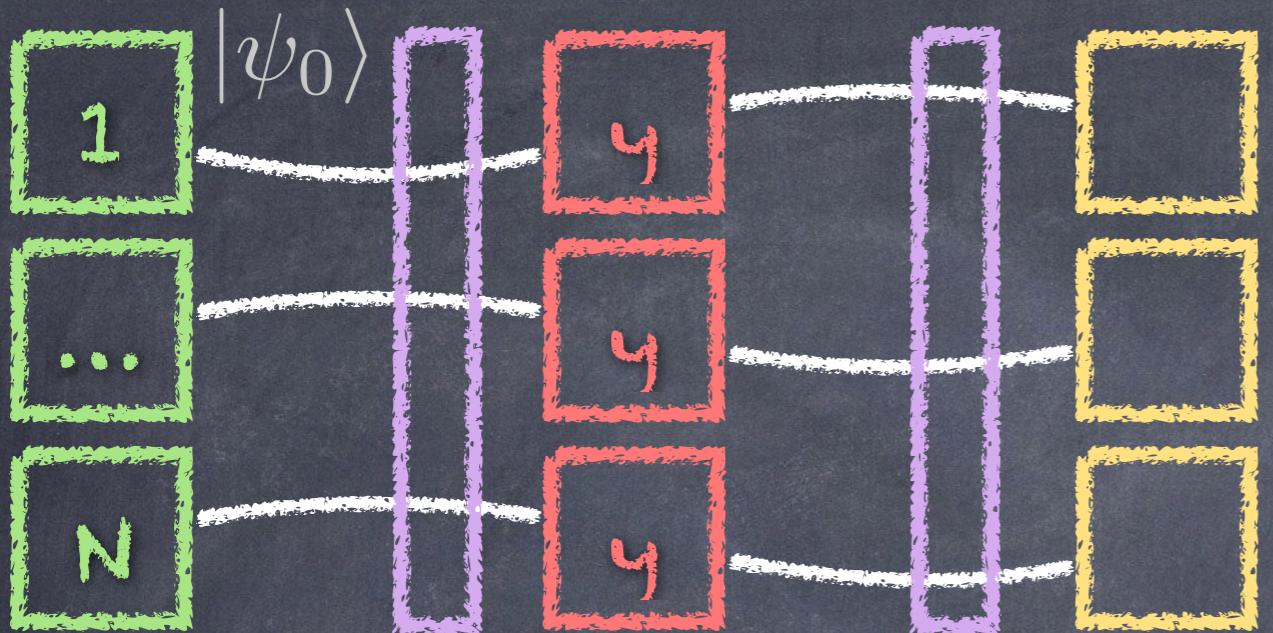
If  $C \ll Q$ , then  $\mathcal{I}_Q(y) \sim N$ .



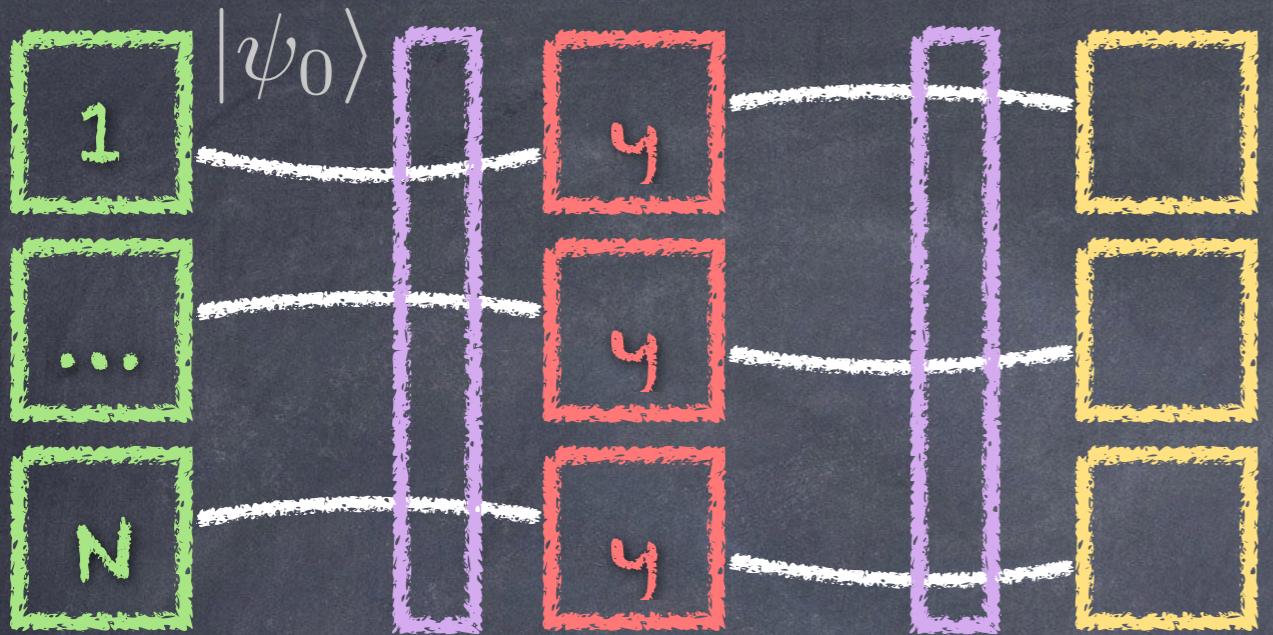
$$U_y = \exp(-iH y)$$

**C**       $\mathcal{I}_Q(y) \sim N$       **Q**       $\mathcal{I}_Q(y) \sim N^2$

**Q**       $\mathcal{I}_Q(y) \sim N$       **Q**       $\mathcal{I}_Q(y) \sim N^2$



$$U_y = \exp(-iH y)$$



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$$\text{Var}[y] \Delta^2 H_{\text{tot}} \geq \frac{1}{4M}$$

generalised Heisenberg relation