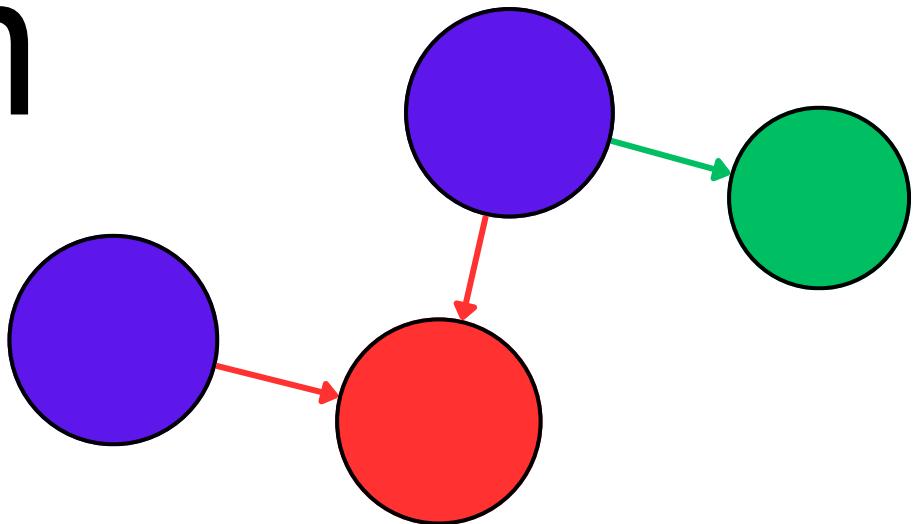
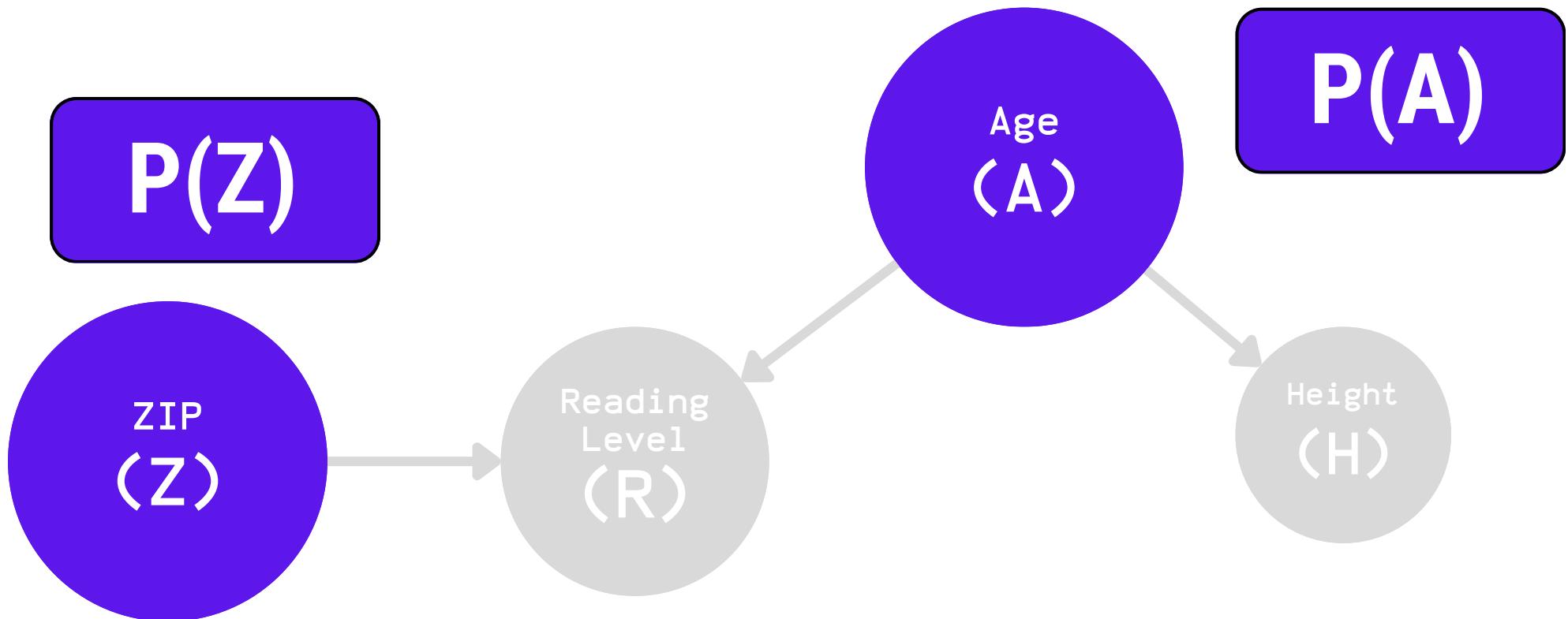


Going from a DAG to a factorization:

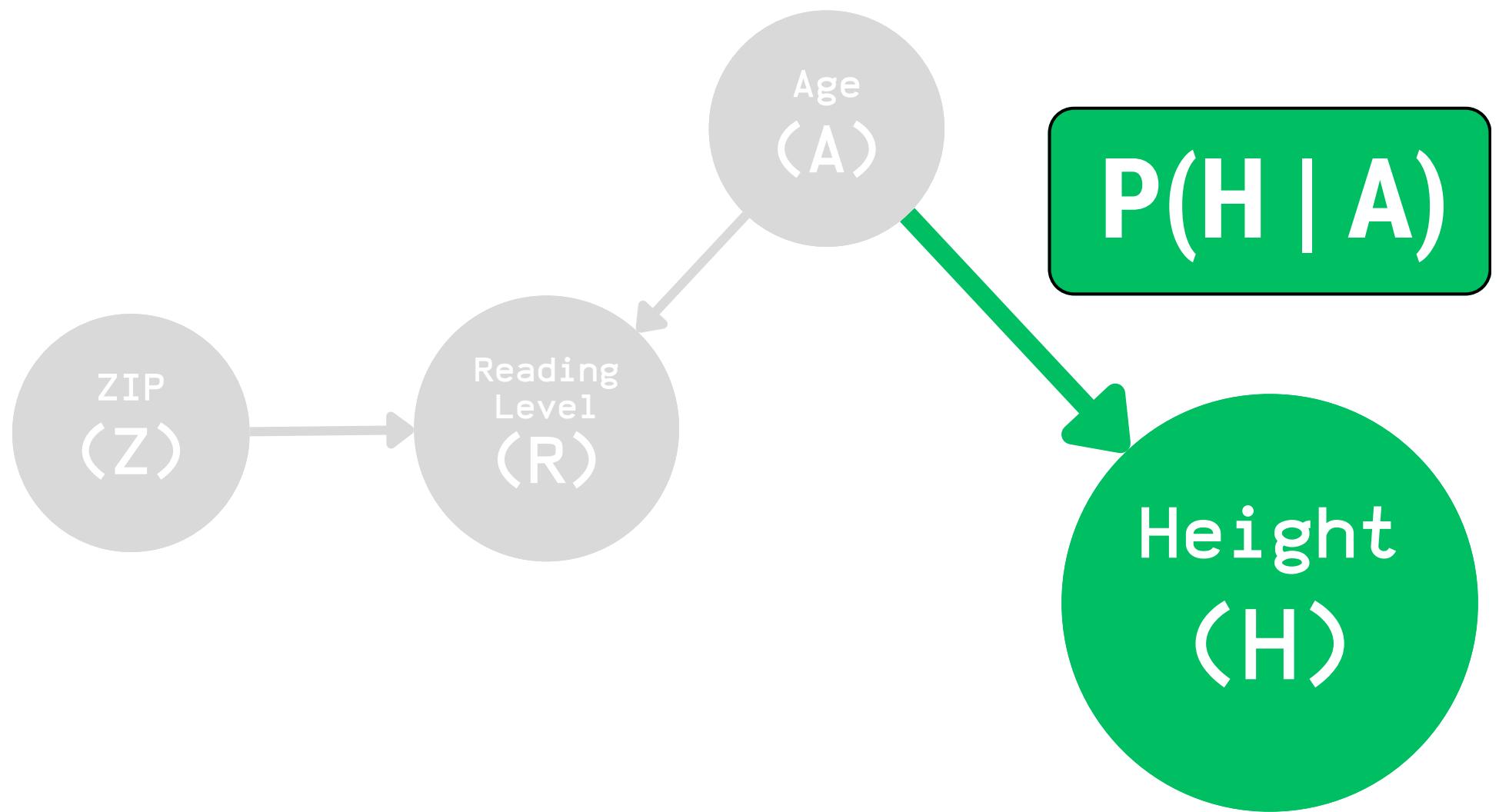
**the “code” in
3 Rules**





When a node has NO arrows pointing into it... we must include each

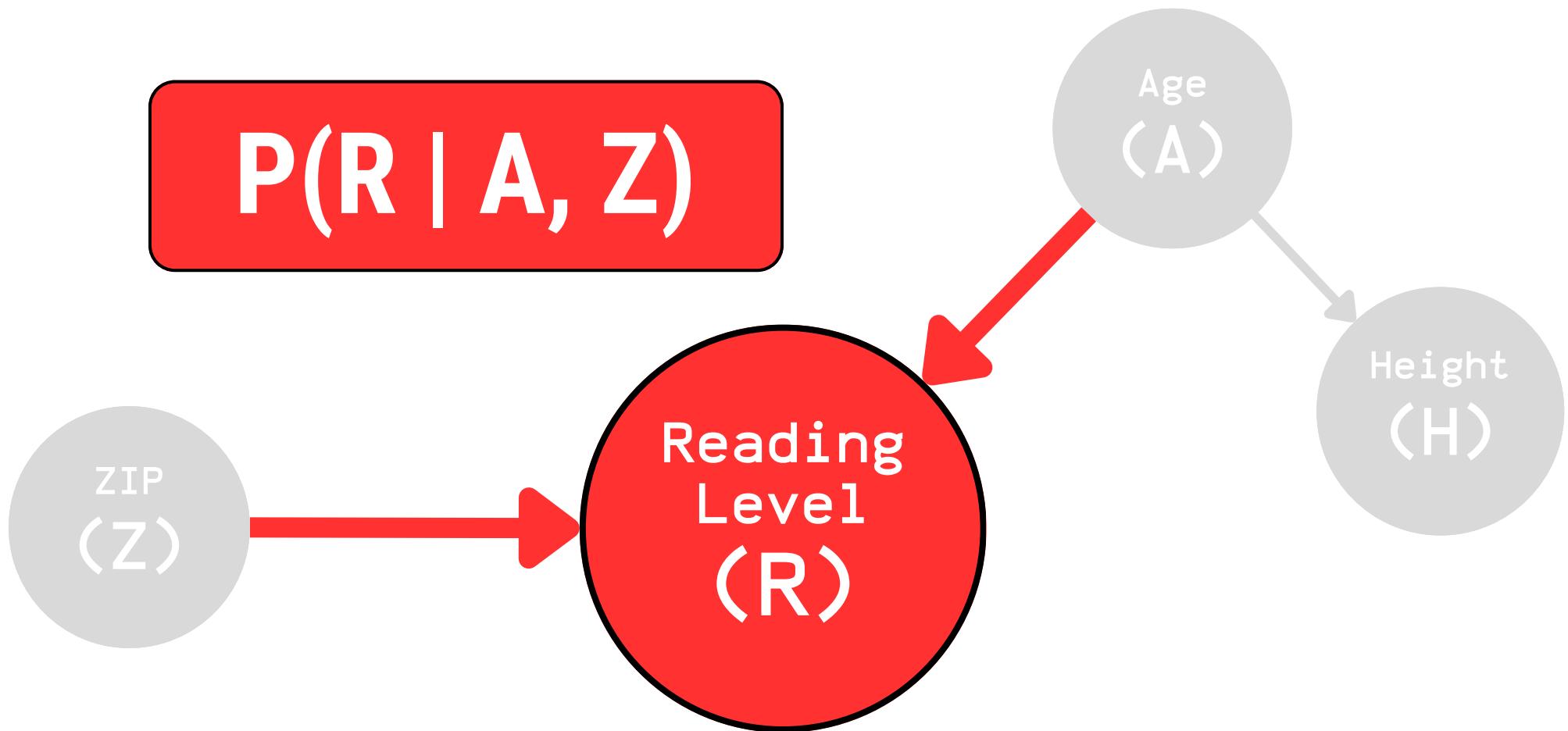
MARGINAL PROBABILITY



When a node has ONLY ONE arrow pointing into it... we must include that

CONDITIONAL PROBABILITY

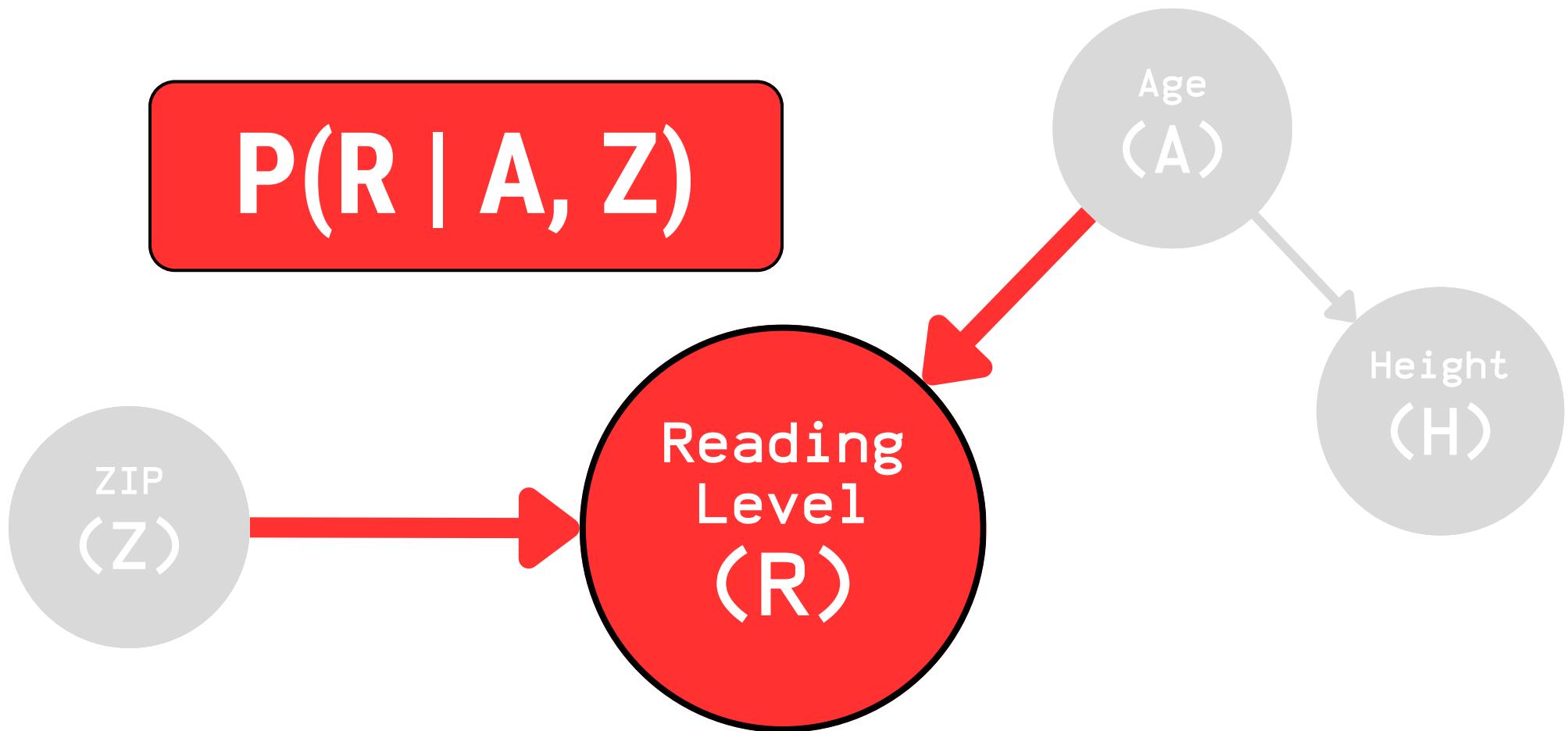
$$P(R | A, Z)$$



*When a node has **MULTIPLE** arrows pointing into it... we must include its*

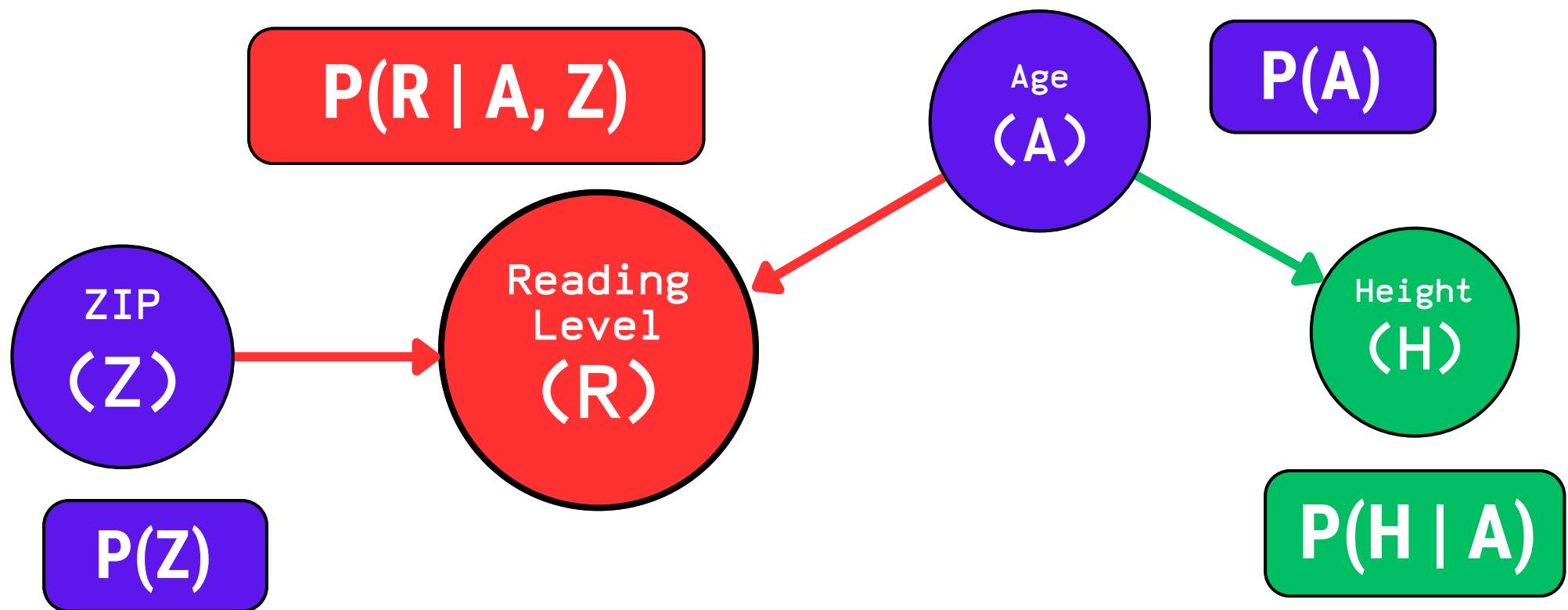
CONDITIONAL PROBABILITY

$$P(R | A, Z)$$



*When a node has **MULTIPLE** arrows pointing into it... we must include its*

CONDITIONAL PROBABILITY



given this DAG... $P(A,H,R,Z)$ factorizes as:

two arrows: ZIP and Age

$P(R | A, Z)$

conditional probability

one arrow: Age

$P(H | A)$

conditional probability

no arrows point at them

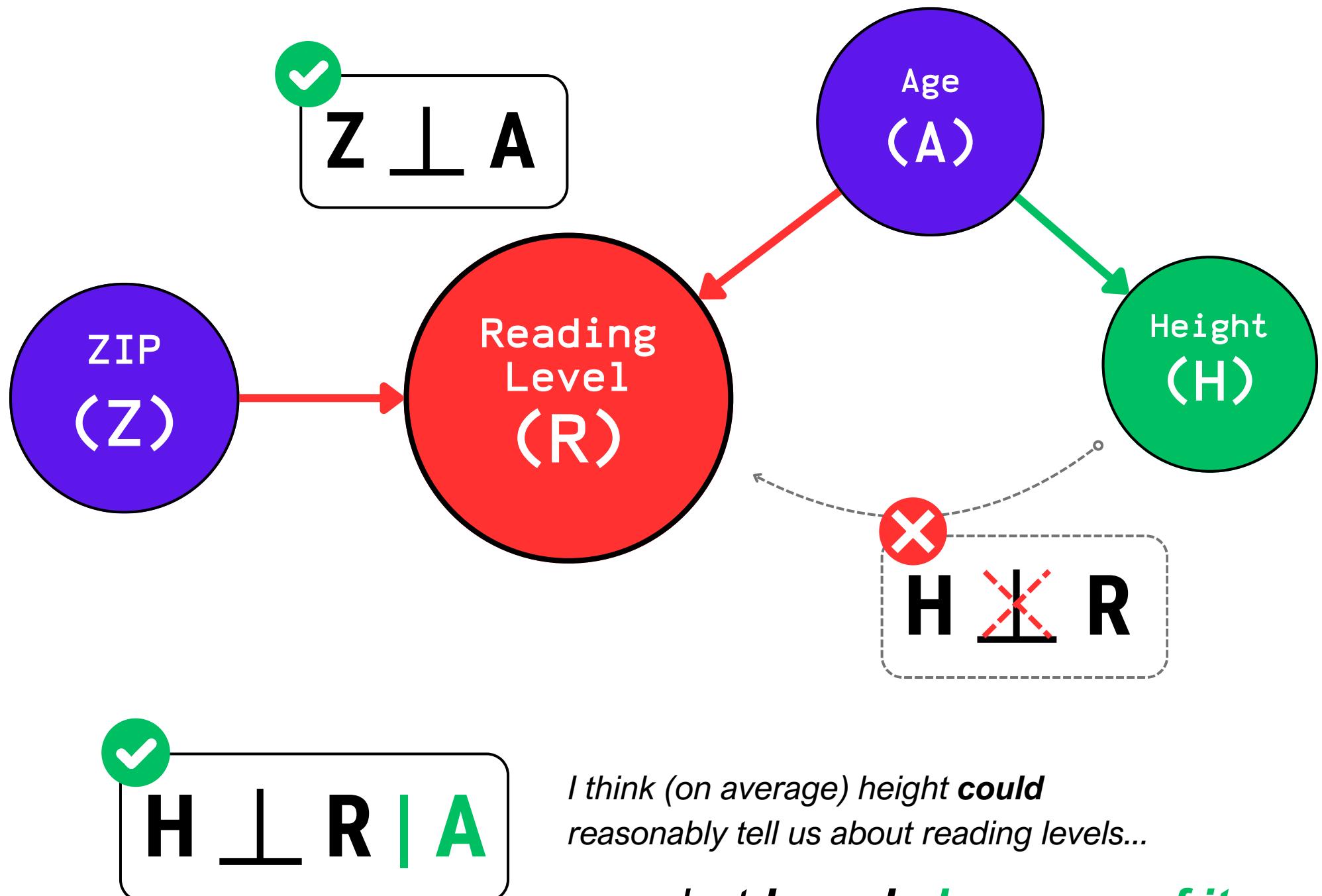
$P(Z)$

marginal probabilities

$P(A)$

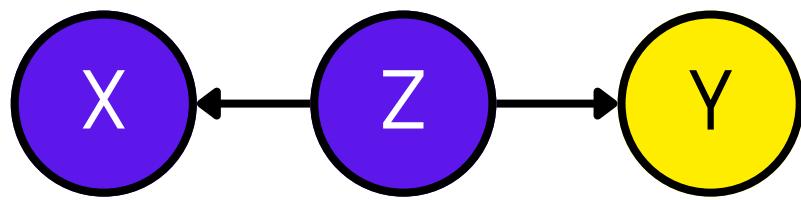
$P(Z)$

$P(H | A)$



*I think (on average) height **could** reasonably tell us about reading levels...*

...but largely because of its relationship with age.



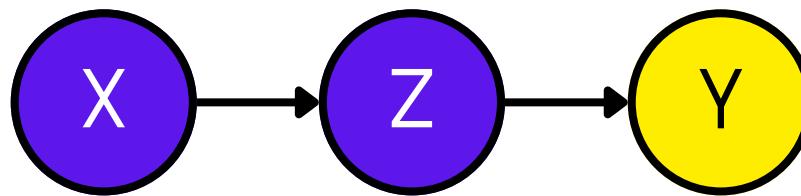
$P(X | Z)$

$P(Z)$

$P(Y | Z)$

$X \perp\!\!\! \perp Y | Z$

$P(X, Z)$

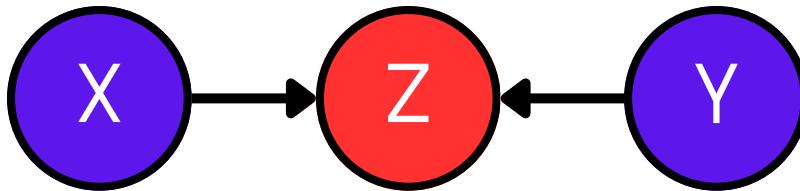


$P(Z | X)$

$P(X)$

$P(Y | Z)$

$X \perp\!\!\! \perp Y$



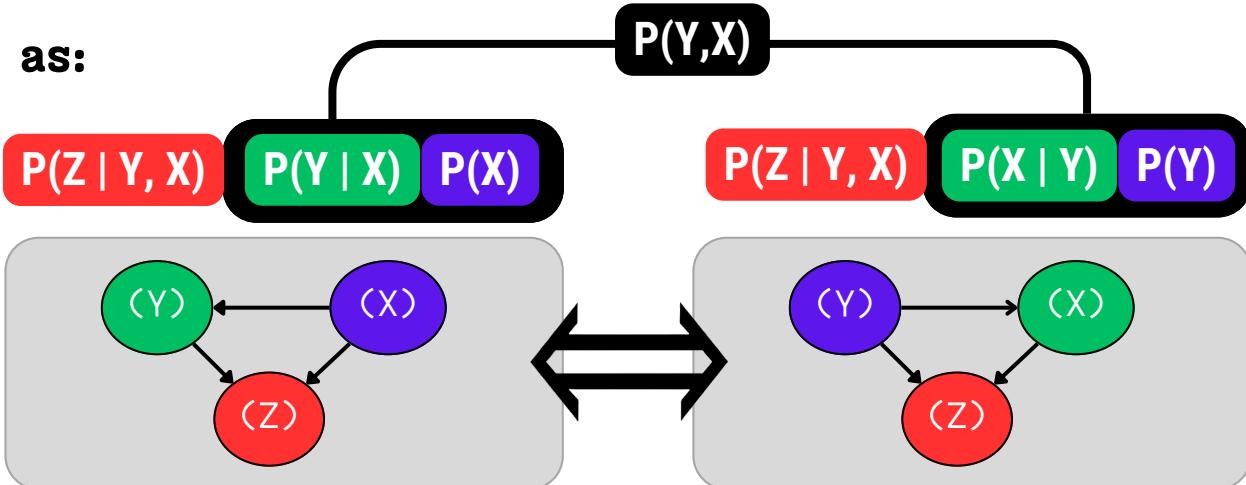
$P(Z | X, Y)$

$P(X)$

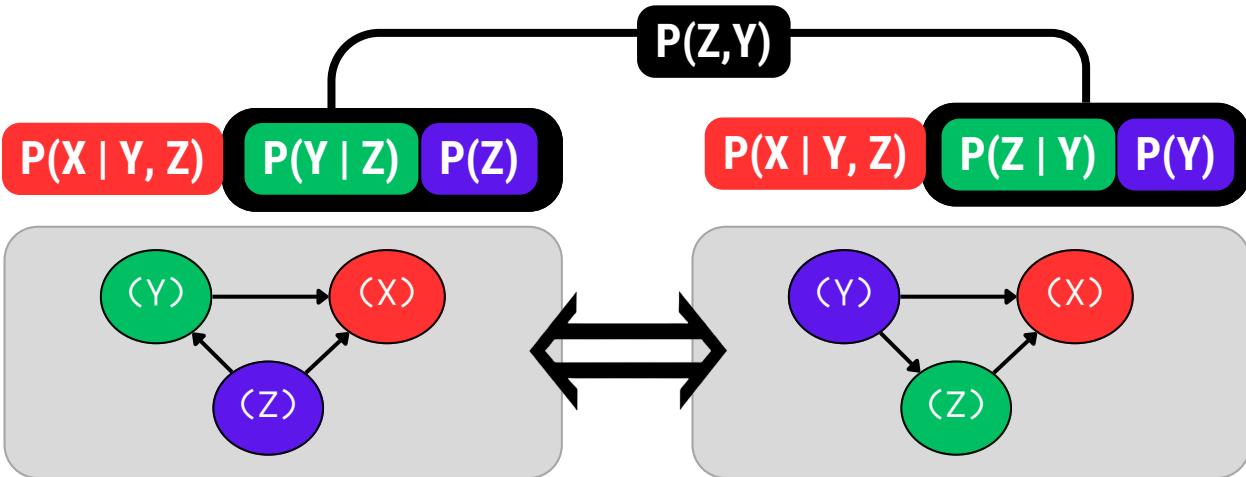
$P(Y)$

Any generic 3-variable joint distribution could be visualized as:

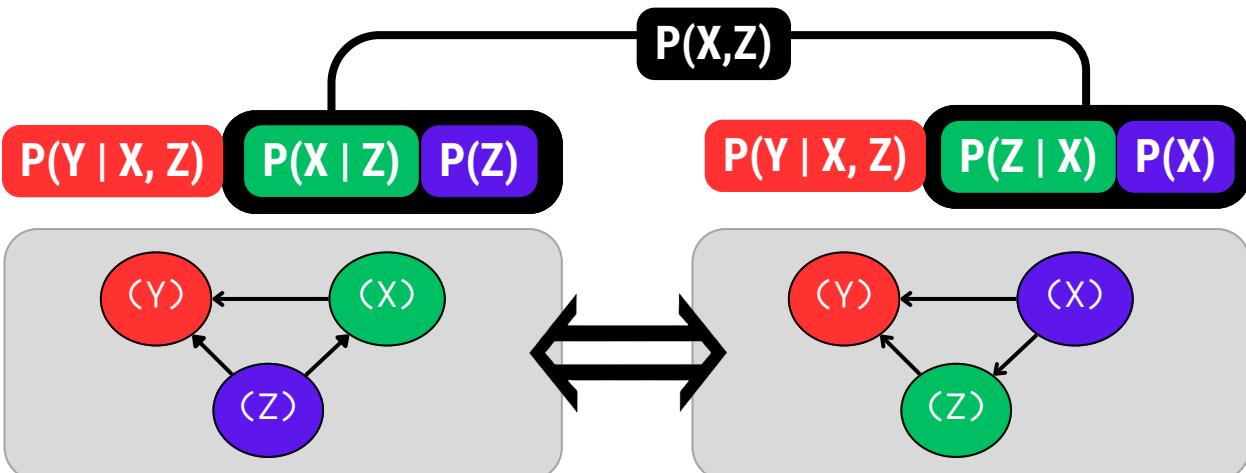
$$P(X, Y, Z) =$$



$$P(Z | Y, X) \quad P(Y, X) =$$



$$P(X | Y, Z) \quad P(Z, Y) =$$



In certain cases, the factorization can simplify to:

$P(Z | Y, X)$

$P(X)$

$P(Y)$

..but **ONLY** when the joint distributions of what we are conditioning on are

INDEPENDENT

$P(Y, X)$



$P(X | Y)$

$P(Y)$

this is
always
true



$P(Y | X)$

$P(X)$

ONLY if this holds



$P(X)$

$P(Y)$

this is **NOT**
always
true

This DAG is telling us
that it does:

$X \perp Y$



=

$P(Z | X, Y)$

$P(X)$

$P(Y)$