

Stat 201: Introduction to Statistics

Standard 16: Probability – Medical
Testing Terminology

Example 2: Probability

	Positive Test (Pos)	Negative Test (Neg)	Total
Has Disease (D)	48	6	54
Disease Free (D^c)	1307	3921	5228
Total	1355	3927	5282

- Let Event...
 - D = Has Disease
 - D^c = Disease Free
 - Pos = Positive Test
 - Neg = Negative Test

Example 2: Probability

	Positive Test (Pos)	Negative Test (Neg)	Total
Has Disease (D)	48	6	54
Disease Free (D^c)	1307	3921	5228
Total	1355	3927	5282

- A **true positive** is when a participant tests positive for the disease and does have it
- Here there are 48 true positives

Example 2: Probability

	Positive Test (Pos)	Negative Test (Neg)	Total
Has Disease (D)	48	6	54
Disease Free (D^c)	1307	3921	5228
Total	1355	3927	5282

- A **true negative** is when a participant tests negative for the disease and doesn't have it
- Here there are 3,921 true negatives

Example 2: Probability

	Positive Test (Pos)	Negative Test (Neg)	Total
Has Disease (D)	48	6	54
Disease Free (D^c)	1307	3921	5228
Total	1355	3927	5282

- A **false positive** is when a participant tests positive for the disease but doesn't have it
- Here there are 1,307 false positives

Example 2: Probability

	Positive Test (Pos)	Negative Test (Neg)	Total
Has Disease (D)	48	6	54
Disease Free (D^c)	1307	3921	5228
Total	1355	3927	5282

- A **false negative** is when a participant tests negative for the disease but does have it
- Here there are 6 false negatives

Example 2: Probability

	Positive Test (Pos)	Negative Test (Neg)	Total
Has Disease (D)	48	6	54
Disease Free (D^c)	1307	3921	5228
Total	1355	3927	5282

- The probability a randomly selected participant had a positive test

$$\begin{aligned} P(Pos) &= \frac{\text{Number of positive observations}}{\text{Total number of observations}} \\ &= \frac{1355}{5282} = .25653162 \end{aligned}$$

Example 2: Probability

	Positive Test (Pos)	Negative Test (Neg)	Total
Has Disease (D)	48	6	54
Disease Free (D^c)	1307	3921	5228
Total	1355	3927	5282

- The probability a randomly selected participant had Disease:

$$\begin{aligned} P(D) &= \frac{\text{Number of } D \text{ observations}}{\text{Total number of observations}} \\ &= \frac{54}{5282} = .0102234 \end{aligned}$$

Example 2: Probability

	Positive Test (Pos)	Negative Test (Neg)	Total
Has Disease (D)	48	6	54
Disease Free (D^c)	1307	3921	5228
Total	1355	3927	5282

- The probability a randomly selected participant had Disease given they tested positive

$$\begin{aligned} P(D|Pos) &= \frac{P(D \& Pos)}{P(Pos)} = \frac{\left(\frac{48}{5282}\right)}{.25653162} \\ &= .03542435 \end{aligned}$$

Example 2: Probability

	Positive Test (Pos)	Negative Test (Neg)	Total
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Disease Free (D^c)	1307	3921	5228
Total	1355	3927	5282

- As before we can rewrite this as:

$$P(D|Pos) = \frac{\text{Number of } D\&Pos \text{ Observations}}{\text{Number of } Pos \text{ Observations}}$$
$$= \frac{48}{1355} = .03542435$$

Example 2: Probability

- $P(D|Pos) = .03542435$
- $P(D) = .0102234$
- Because $P(D|Pos) \neq P(D)$ events D and POS are **not independent** events

Adjectives for Tests

- **Sensitivity** – Probability that a test detects a substance correctly by giving a positive test result
 - $Sensitivity = P(Pos|Substance)$
- **Specificity** – Probability that a test correctly does not detect a substance by giving a negative result
 - $Specificity = P(Neg|Substance^c)$

Example 3: Adjectives for Tests

	Positive Test (Pos)	Negative Test (Neg)	Total
Has Disease (D)	48	6	54
Disease Free (D^c)	1307	3921	5228
Total	1355	3927	5282

- $$\text{Sensitivity} = P(\text{Pos}|D) = \frac{P(\text{Pos} \& D)}{P(D)} =$$
$$\frac{\left(\frac{48}{5282}\right)}{\left(\frac{54}{5282}\right)} = \frac{48}{54} = .8889$$

Example 3: Adjectives for Tests

	Positive Test (Pos)	Negative Test (Neg)	Total
Has Disease (D)	48	6	54
Disease Free (D^c)	1307	3921	5228
Total	1355	3927	5282

- As before we can rewrite this as:

$$\textit{Sensitivity} = \frac{\textit{number of Pos\&D observations}}{\textit{number of D observations}} =$$

$$\frac{48}{54} = .8889$$

Example 3: Adjectives for Tests

	Positive Test (Pos)	Negative Test (Neg)	Total
Has Disease (D)	48	6	54
Disease Free (D^c)	1307	3921	5228
Total	1355	3927	5282

- $Specificity = P(Neg|D^c) = \frac{P(Neg \& D^c)}{P(D^c)} =$
 $\frac{\left(\frac{3921}{5282}\right)}{\left(\frac{5228}{5282}\right)} = \frac{3921}{5228} = .75$

Example 3: Adjectives for Tests

	Positive Test (Pos)	Negative Test (Neg)	Total
Has Disease (D)	48	6	54
Disease Free (D^c)	1307	3921	5228
Total	1355	3927	5282

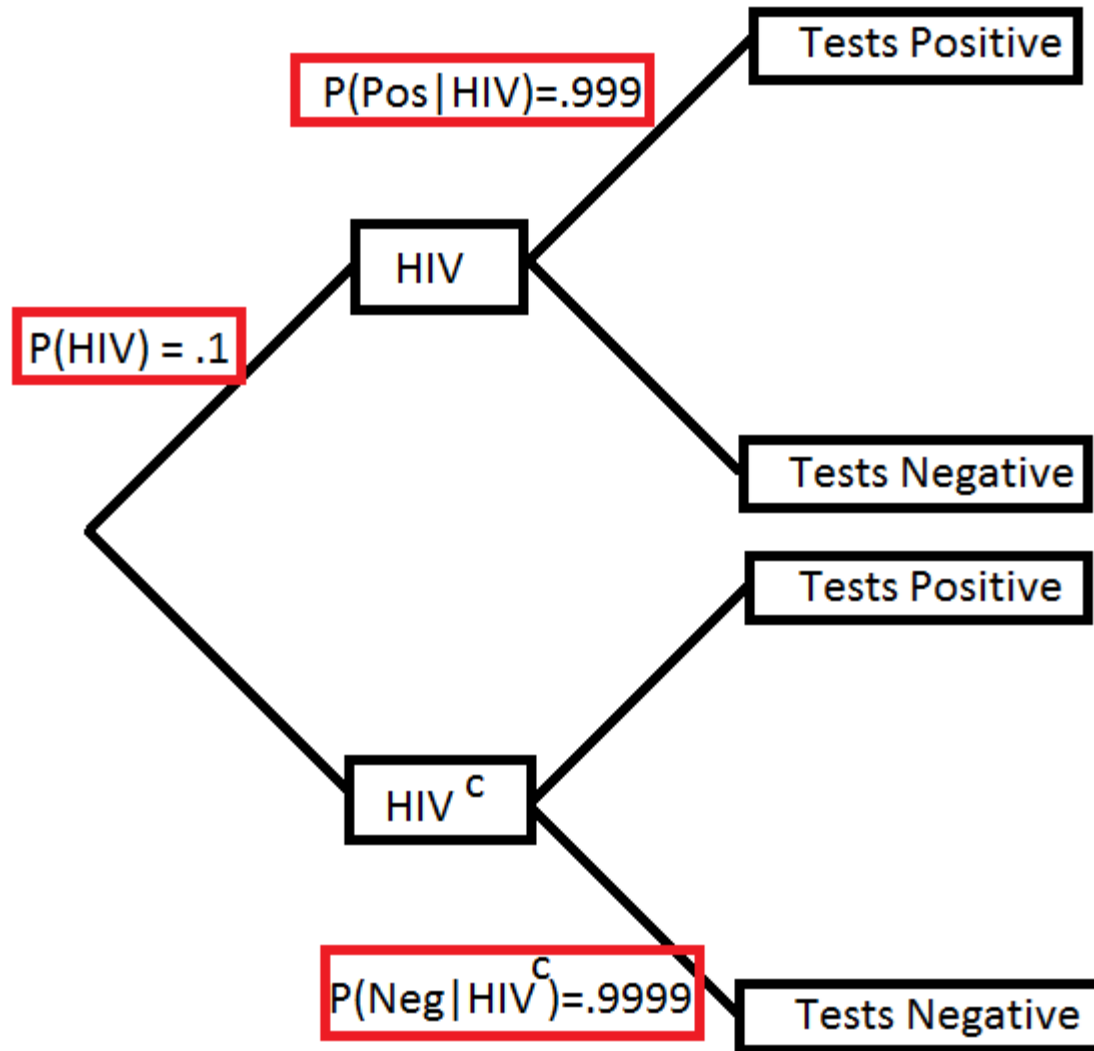
- As before we can rewrite this as:

$$\begin{aligned} \text{Specificity} &= \frac{\text{number of Neg\&D}^c \text{ observations}}{\text{number of D}^c \text{ observations}} \\ &= \frac{3921}{5228} = .75 \end{aligned}$$

Example 4: Tree Diagrams

- For a Western blot blood test the sensitivity is about .999 and the specificity is about .9999
 - *Sensitivity* = $P(Pos|HIV) = .999$
 - *Specificity* = $P(Neg|HIV^c) = .999$
- Consider a high risk group here 10% are truly HIV positive
 - $P(HIV) = .1$

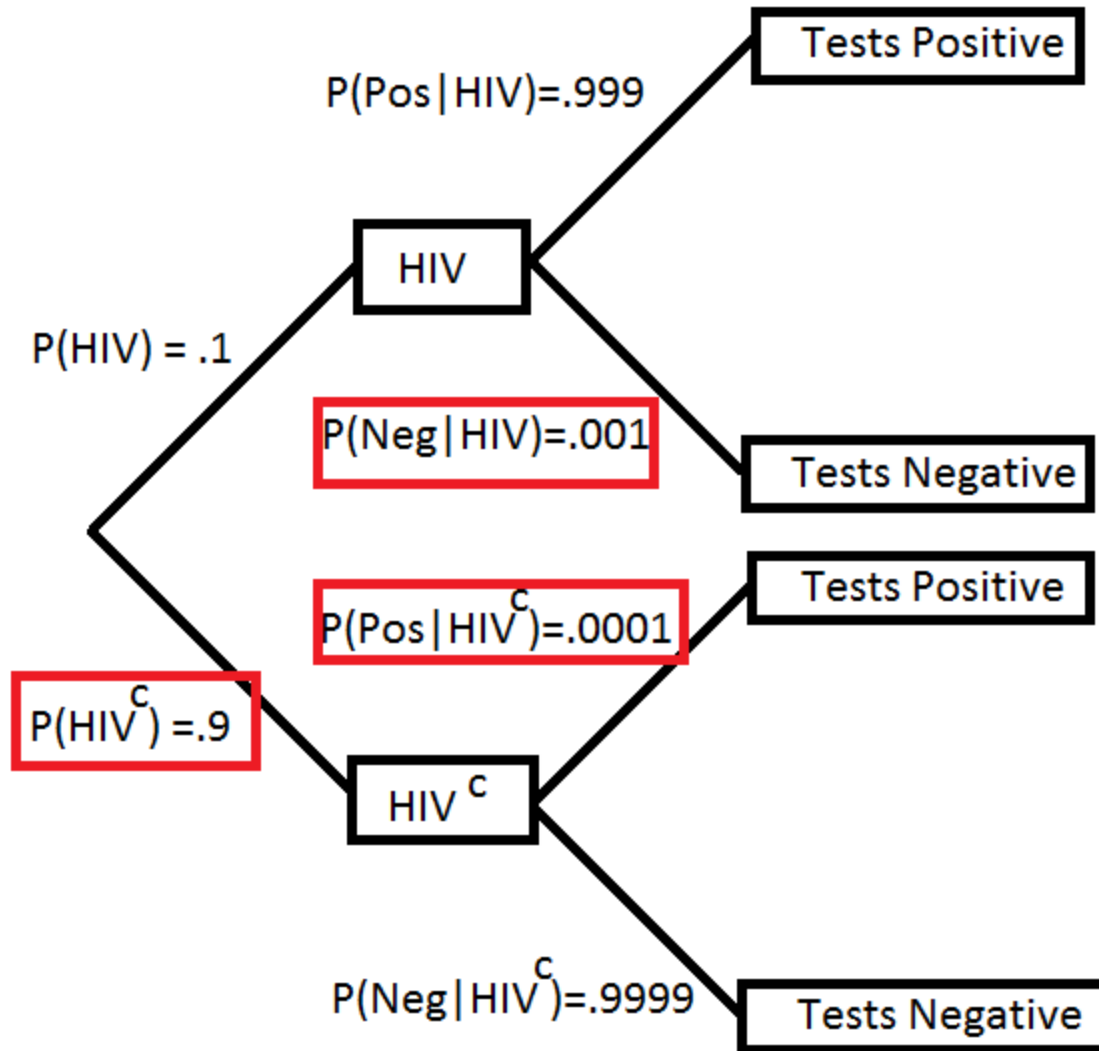
Example 4: Tree Diagrams



Example 4: Tree Diagrams

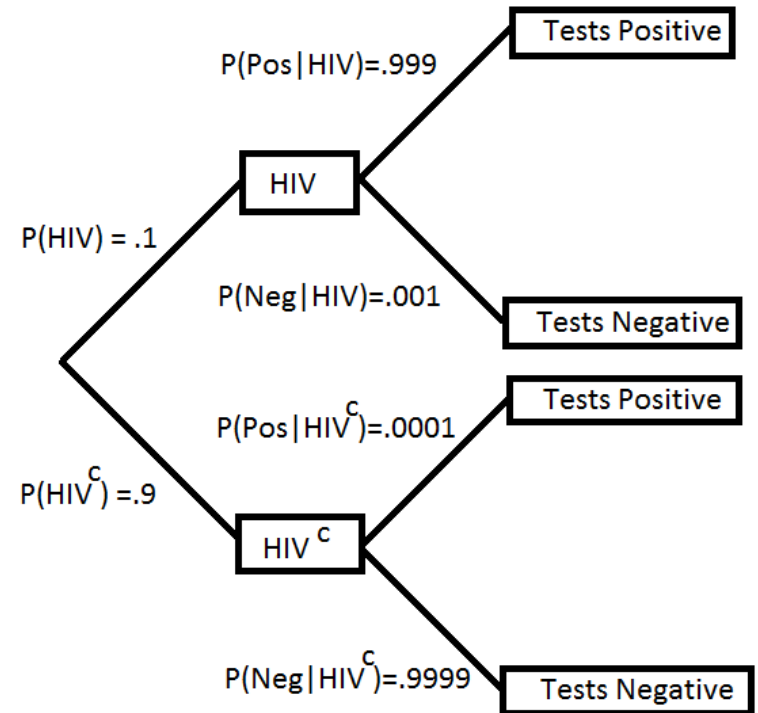
- Using Complement Rule we can find:
 - $P(Neg|HIV) = P(Pos^c|HIV) = 1 - P(Pos|HIV) = 1 - .999 = .001$
 - $P(Pos|HIV^c) = P(Neg^c|HIV^c) = 1 - P(Neg|HIV^c) = 1 - .9999 = .0001$
 - $P(HIV^c) = 1 - P(HIV) = 1 - .1 = .9$

Example 4: Tree Diagrams



Example 4: Tree Diagrams

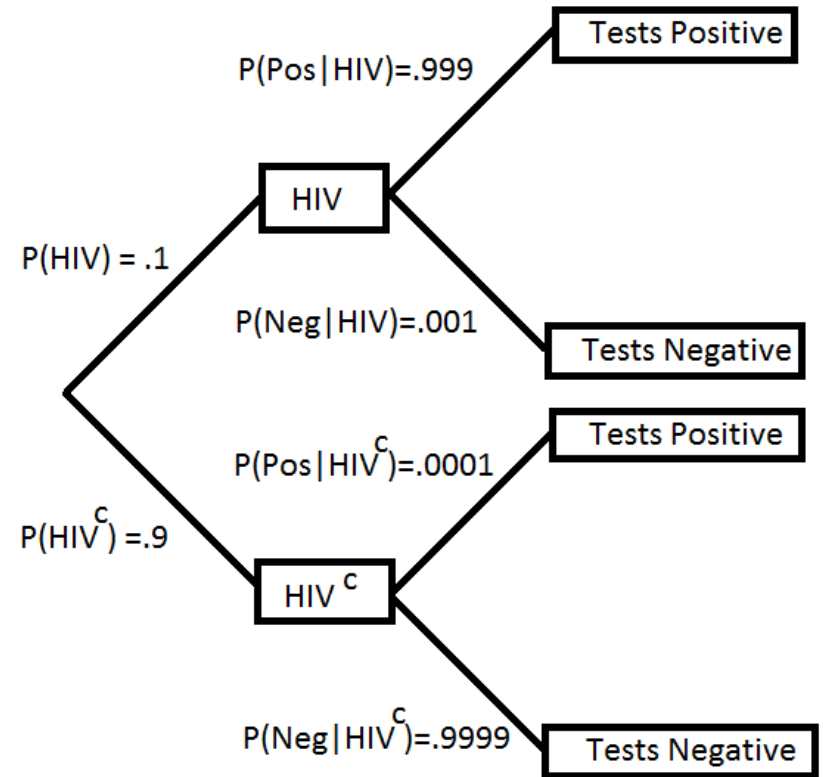
- Note, using the complement rule helped us find the rest of the probabilities
- The branches leaving any point should have probabilities that add to one



Example 4: Tree Diagrams

- The probability that a randomly selected participant tests positive **and** has HIV:

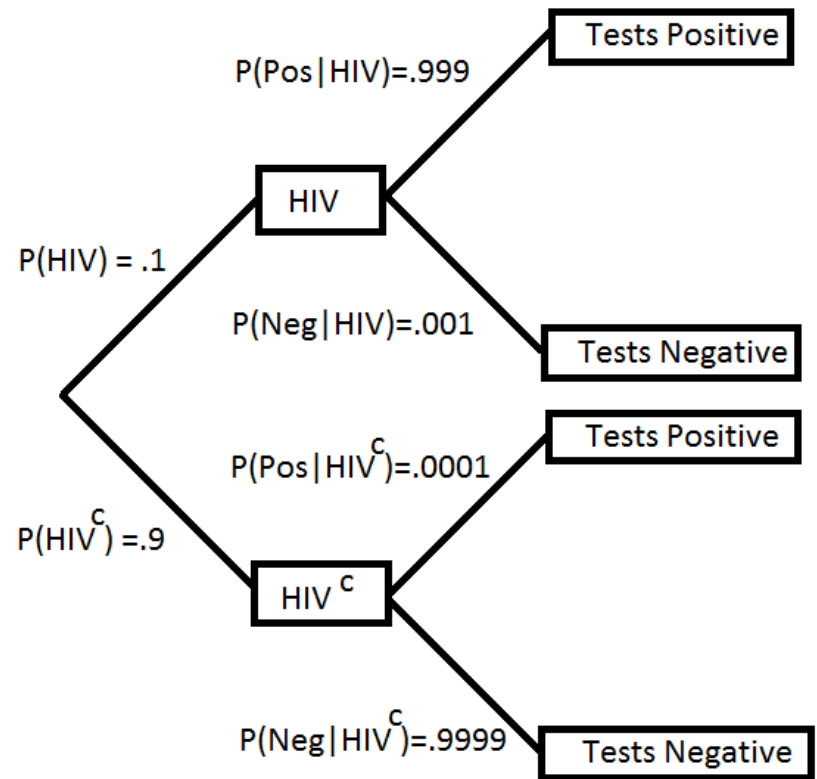
$$\begin{aligned} P(HIV \cap Pos) \\ &= P(HIV) * P(Pos|HIV) \\ &= .1 * .999 = .0999 \end{aligned}$$



Example 4: Tree Diagrams

- The probability that a randomly selected participant tests negative **and** has HIV:

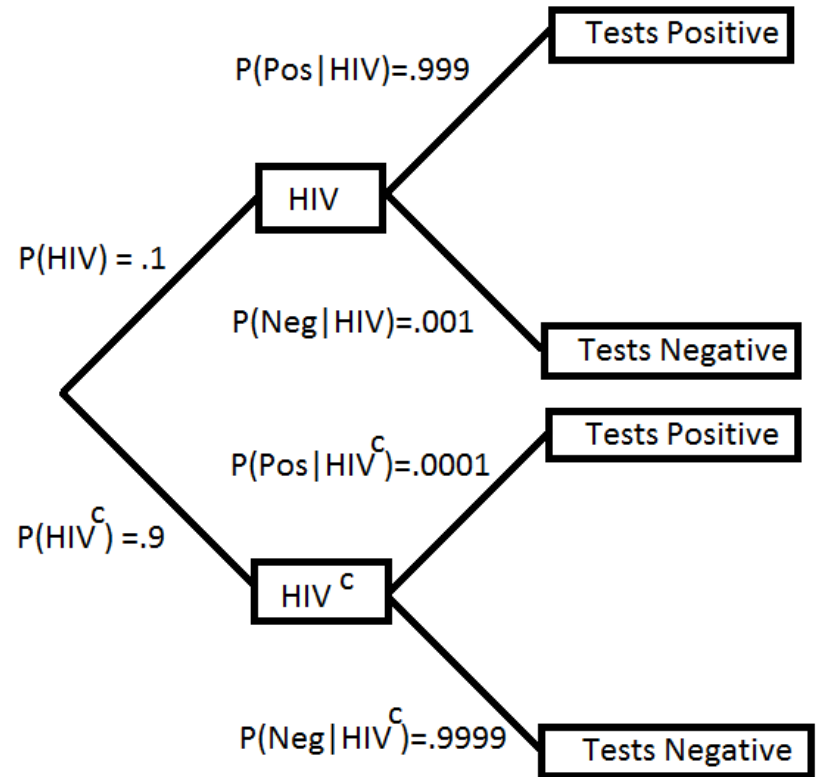
$$\begin{aligned} P(HIV \cap Neg) &= P(HIV) \\ &\quad * P(Neg|HIV) \\ &= .1 * .001 \\ &= .0001 \end{aligned}$$



Example 4: Tree Diagrams

- The probability that a randomly selected participant tests positive **and** doesn't have HIV:

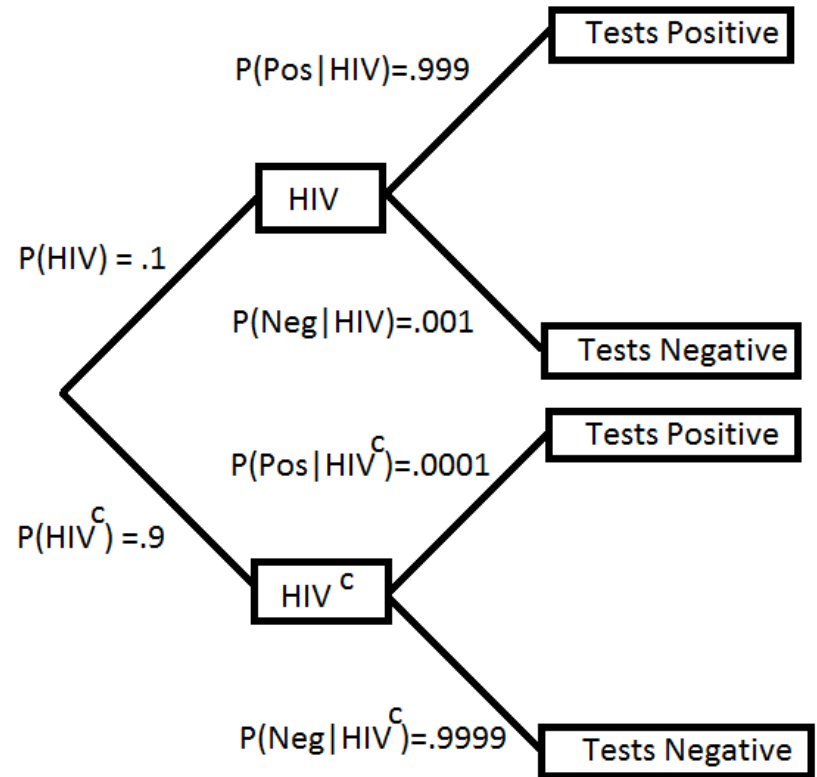
$$\begin{aligned} P\left(HIV^c \cap Pos\right) &= P(HIV^c) \\ &* P(Pos|HIV^c) \\ &= .9 * .0001 \\ &= .00009 \end{aligned}$$



Example 4: Tree Diagrams

- The probability that a randomly selected participant tests negative **and** doesn't have HIV:

$$\begin{aligned} P\left(HIV^c \cap Neg\right) &= P(HIV^c) \\ &* P(Neg|HIV^c) \\ &= .9 * .9999 \\ &= .89991 \end{aligned}$$



Example 4: True Positive

- The probability a randomly selected person has HIV given they tested positive

$$\begin{aligned} \blacksquare P(HIV|Pos) &= \frac{P(HIV \cap Pos)}{P(Pos)} = \\ &= \frac{P(HIV \cap Pos)}{P(Pos|HIV^c)P(HIV^c) + P(Pos|HIV)P(HIV)} = \\ &= \frac{.0999}{.0001*.9 + .999*.1} = .999 \end{aligned}$$

- Since this is very close to one, if you test positive, it is very likely that you have HIV

Example 4: True Positive

- The probability a randomly selected person has HIV given they tested positive
 - $$P(HIV|Neg) = \frac{P(HIV \cap Neg)}{P(Neg)} = \frac{.0001}{.9} = .0001$$
- Since this is very close to zero, if you test negative, it is very unlikely that you have HIV