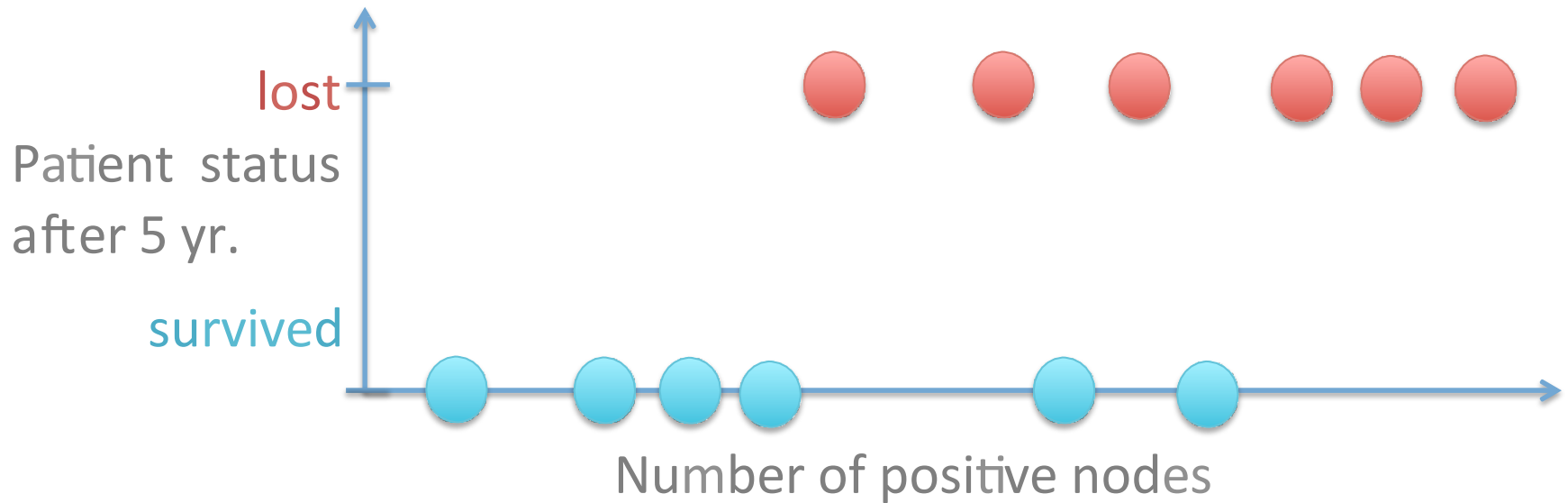


Logistic Regression



1 Feature: Number of + nodes

2 Labels: Survived / Lost



1 Feature: Number of + nodes

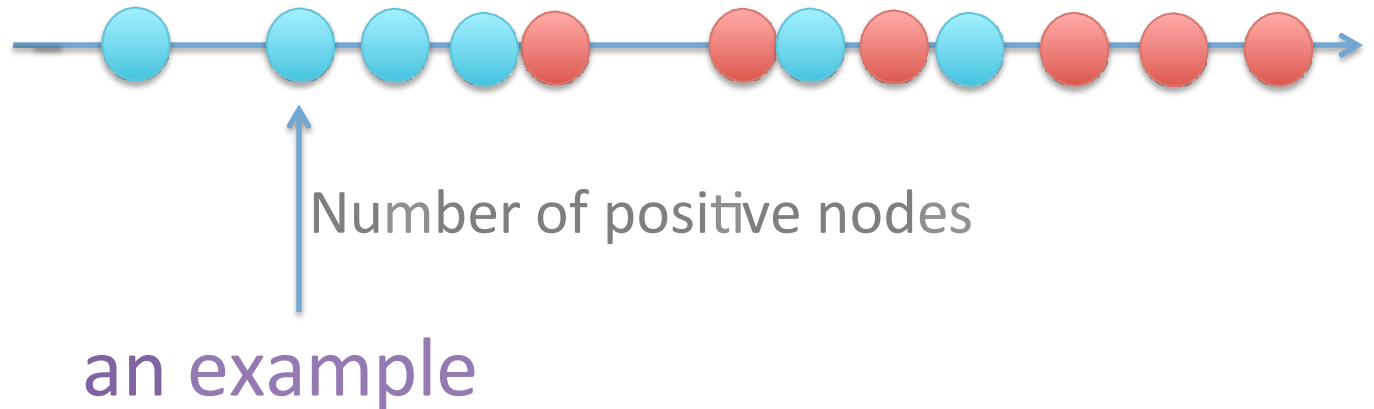
2 Labels: Survived / Lost



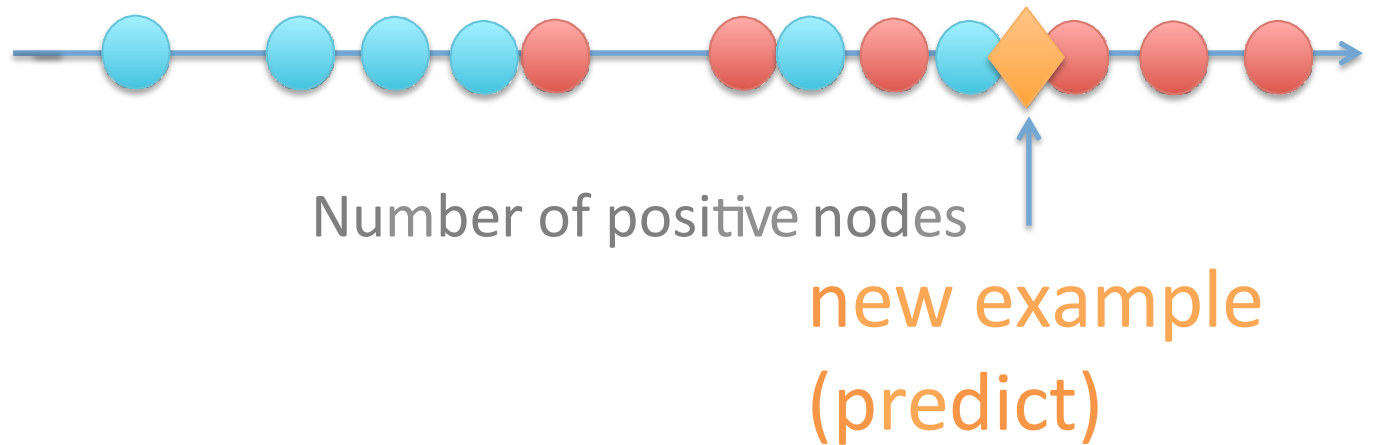
Number of positive nodes

1 Feature: Number of + nodes

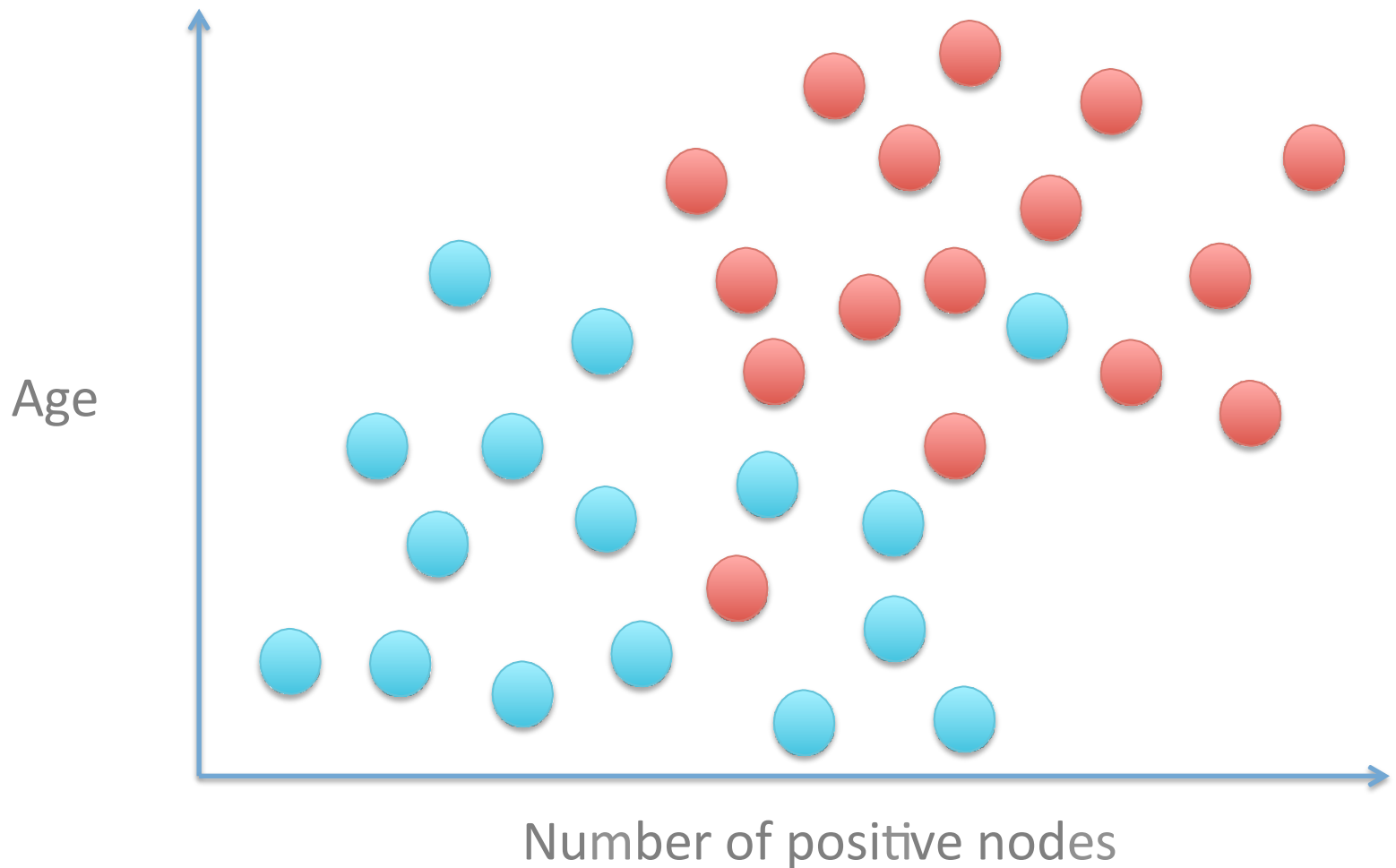
2 Labels: Survived / Lost



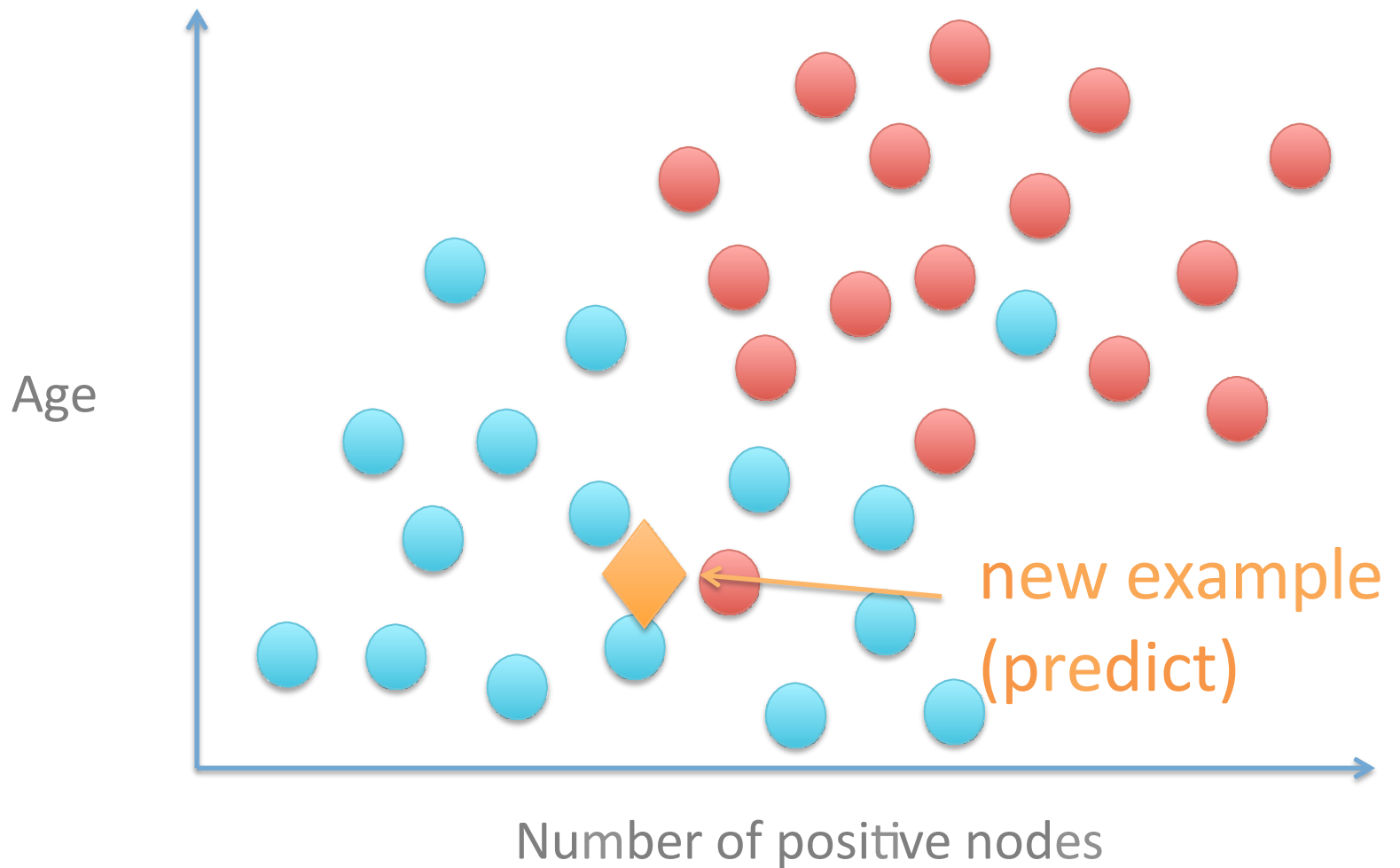
- 1 Feature: Number of + nodes
- 2 Labels: Survived / Lost



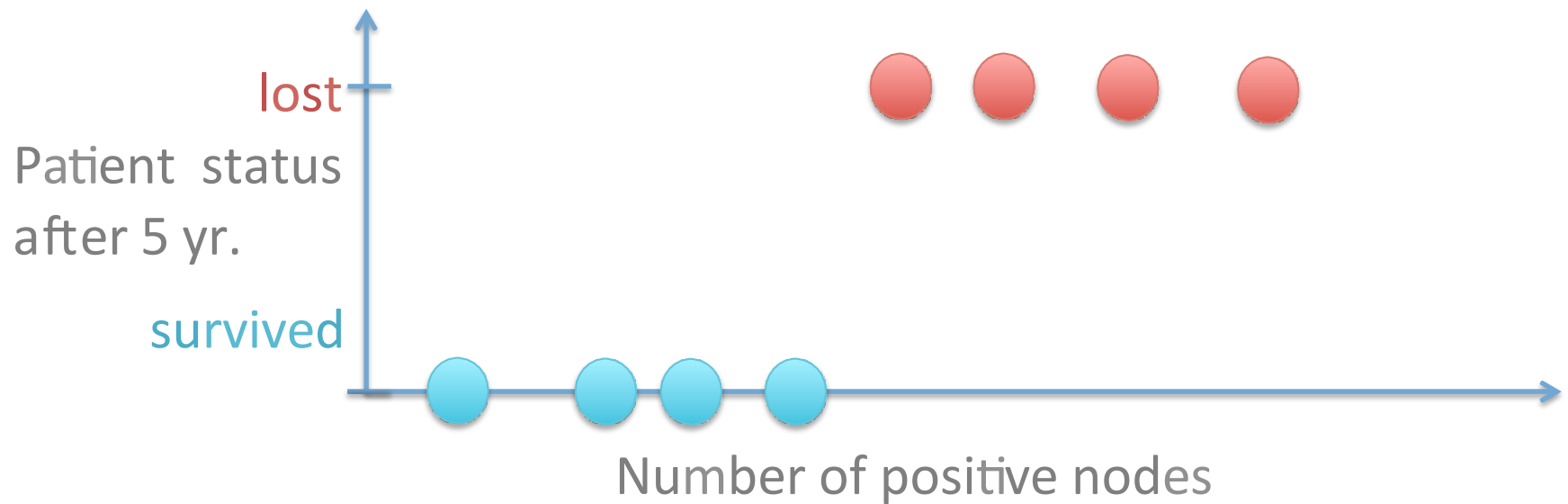
2 Features: Number of + nodes, Age
2 Labels: Survived / Lost



2 Labels: Survived / Lost



Linear regression for classification?

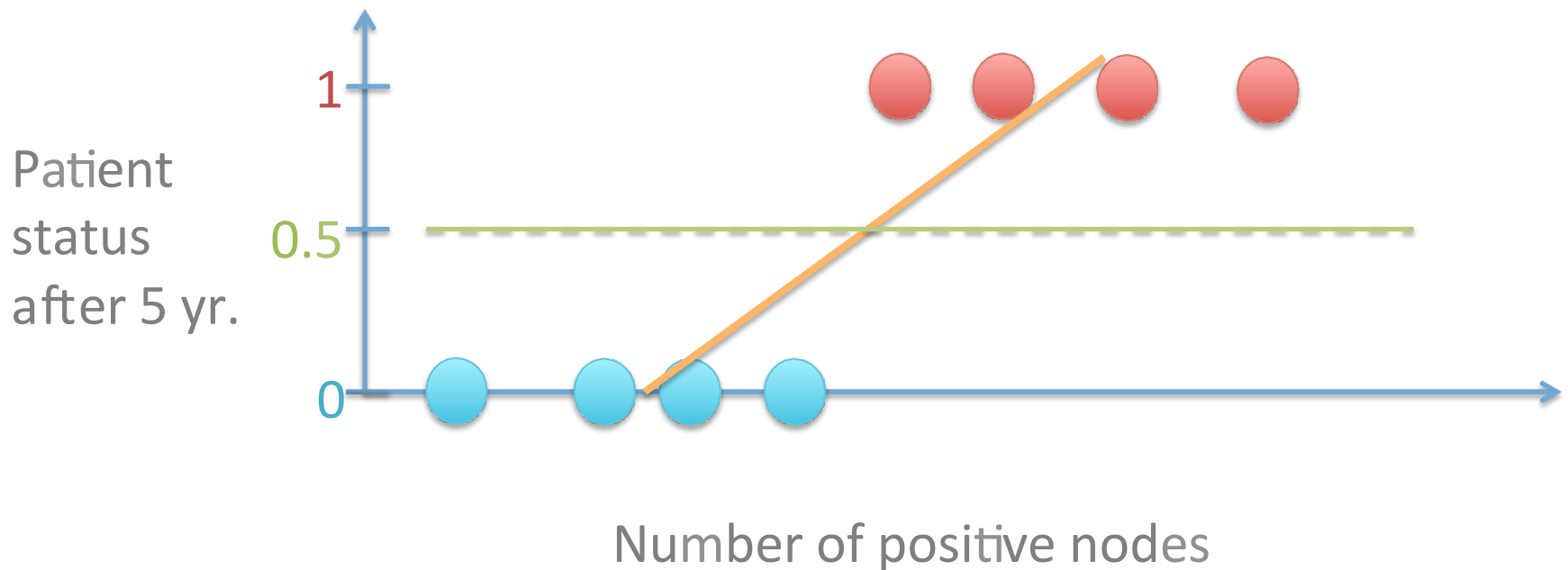


Linear regression for classification?



$$y_{\beta}(x) = \beta_0 + \beta_1 x + \varepsilon$$

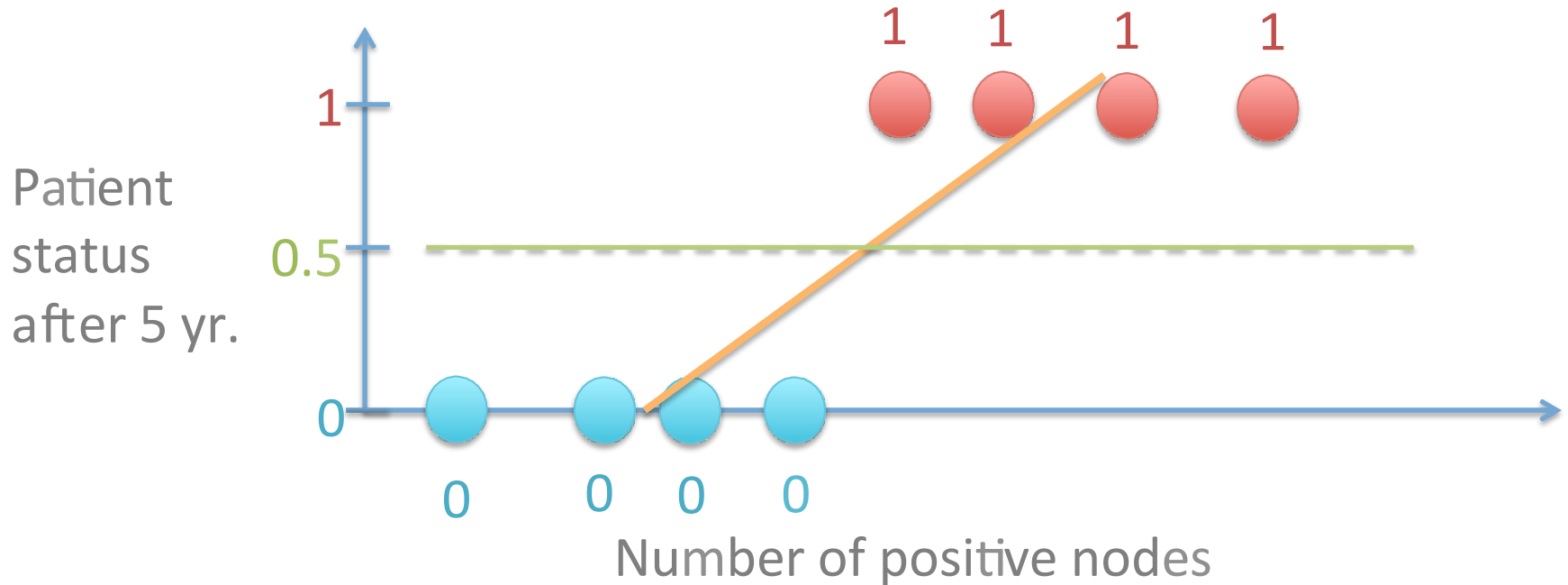
Linear regression for classification?



If $y_{\text{pred}} > 0.5$ predict label 1 (lost)

If $y_{\text{pred}} < 0.5$ predict label 0 (survived)

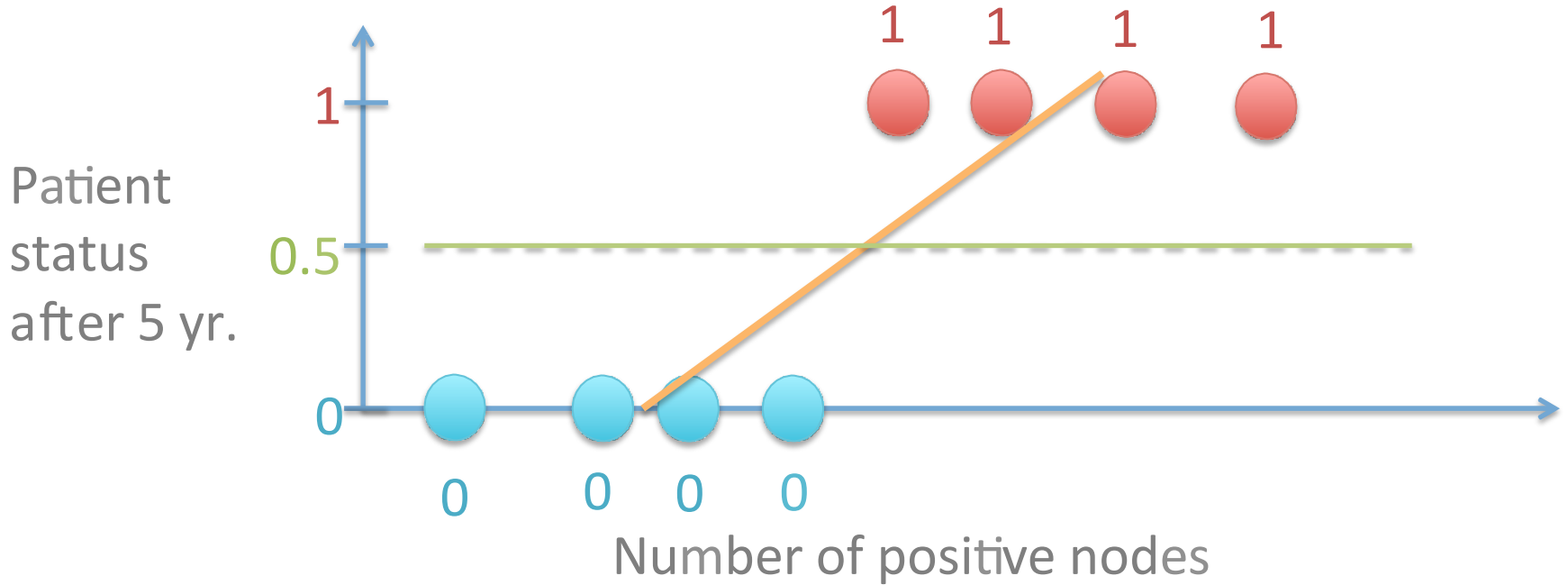
Linear regression for classification?



If $y_{\text{pred}} > 0.5$ predict label 1 (lost)

If $y_{\text{pred}} < 0.5$ predict label 0 (survived)

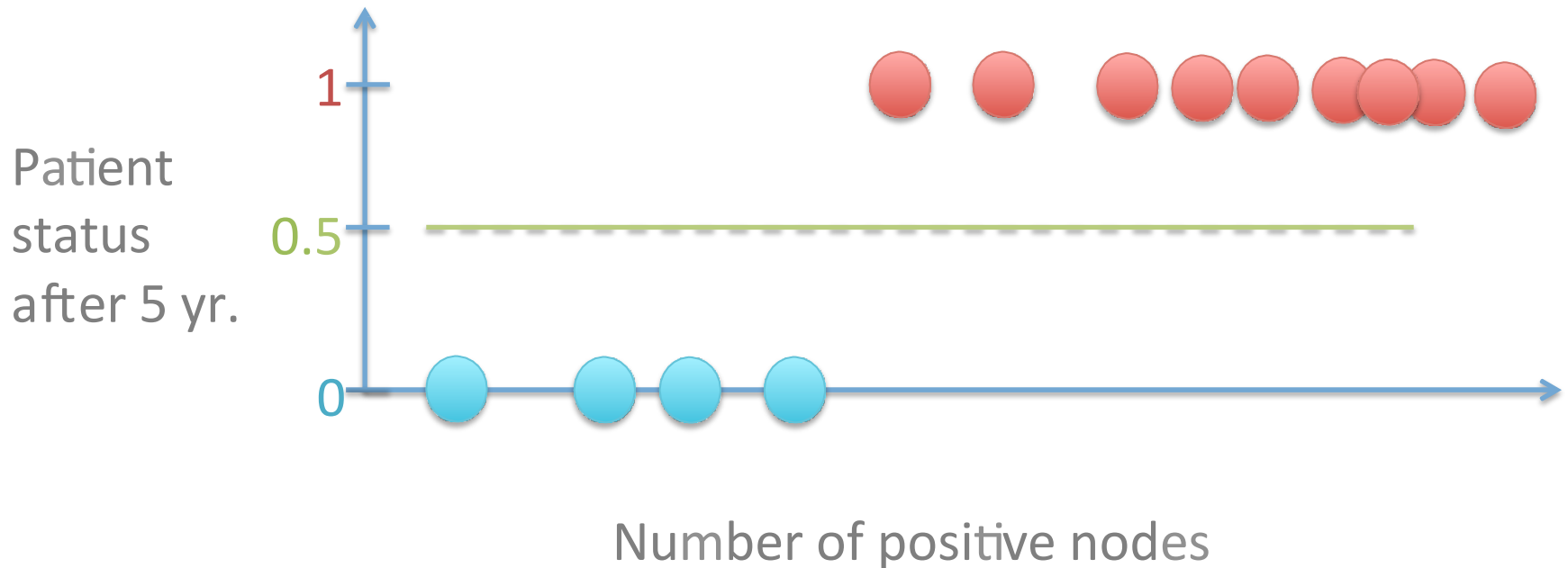
Linear regression for classification?



If $y_{\text{pred}} > 0.5$ predict label 1 (lost)

If $y_{\text{pred}} < 0.5$ predict label 0 (survived)

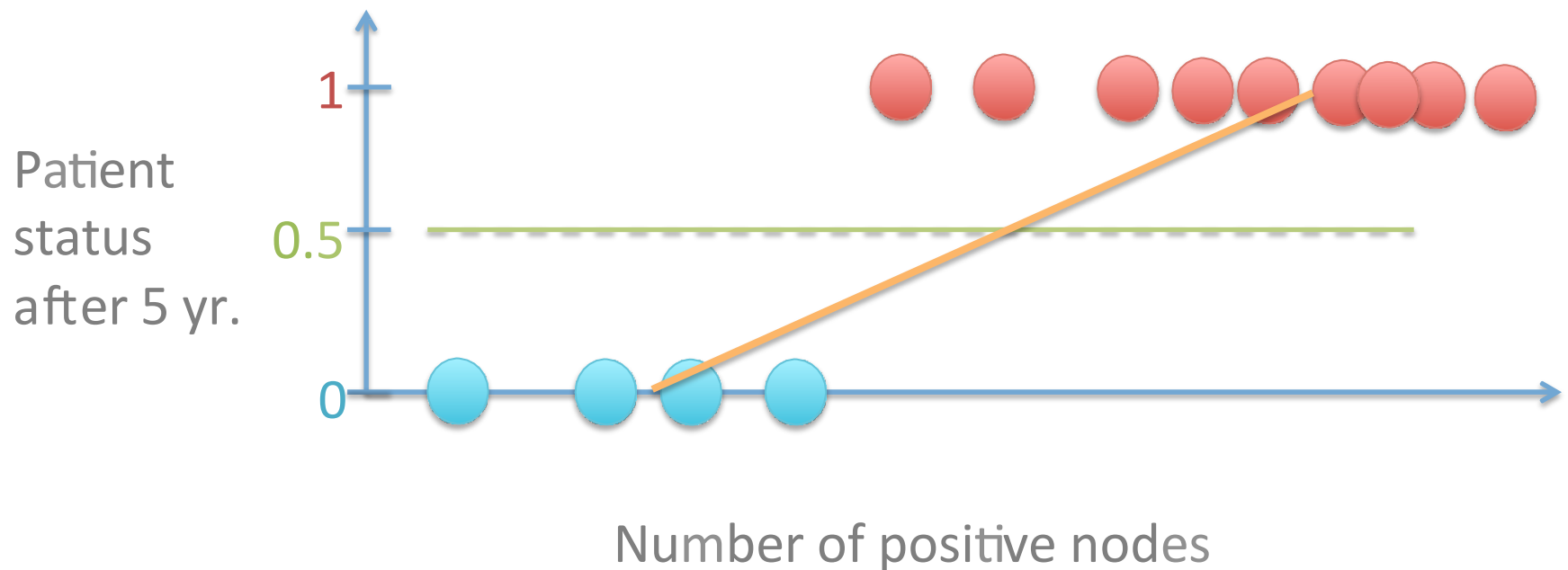
Linear regression for classification?



If $y_{\text{pred}} > 0.5$ predict label 1 (lost)

If $y_{\text{pred}} < 0.5$ predict label 0 (survived)

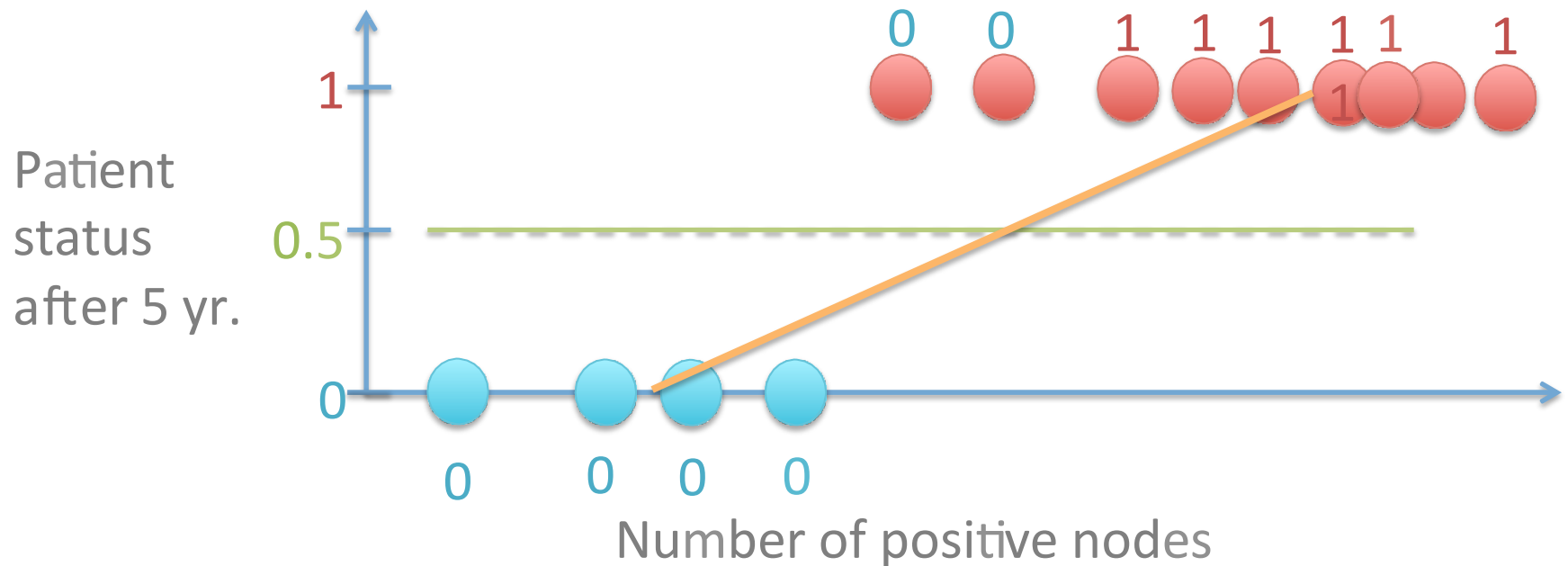
Linear regression for classification?



If $y_{\text{pred}} > 0.5$ predict label 1 (lost)

If $y_{\text{pred}} < 0.5$ predict label 0 (survived)

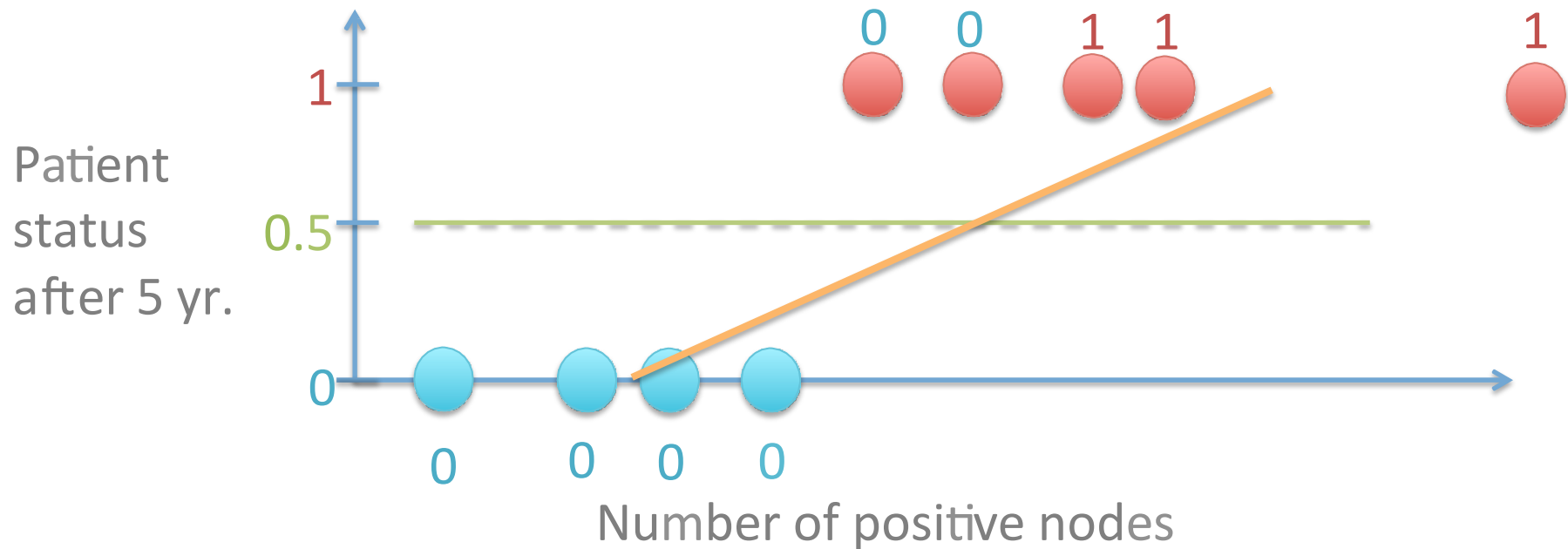
Linear regression for classification?



If $y_{\text{pred}} > 0.5$ predict label 1 (lost)

If $y_{\text{pred}} < 0.5$ predict label 0 (survived)

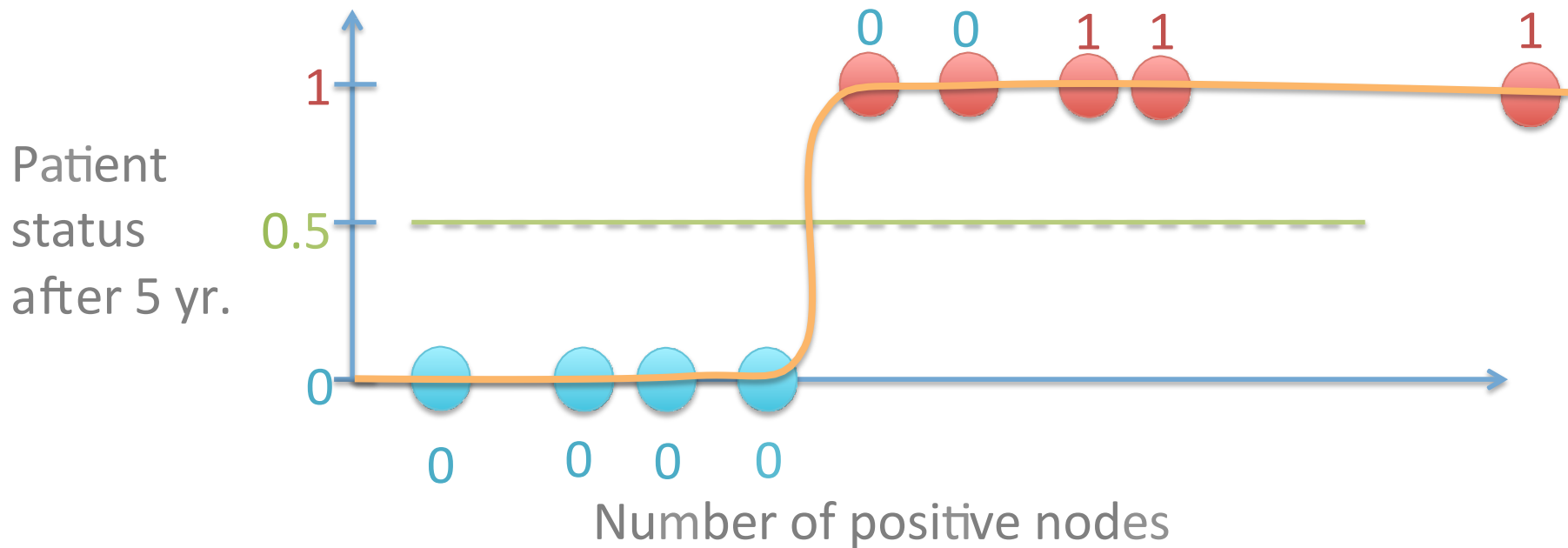
Linear regression for classification?



If $y_{\text{pred}} > 0.5$ predict label 1 (lost)

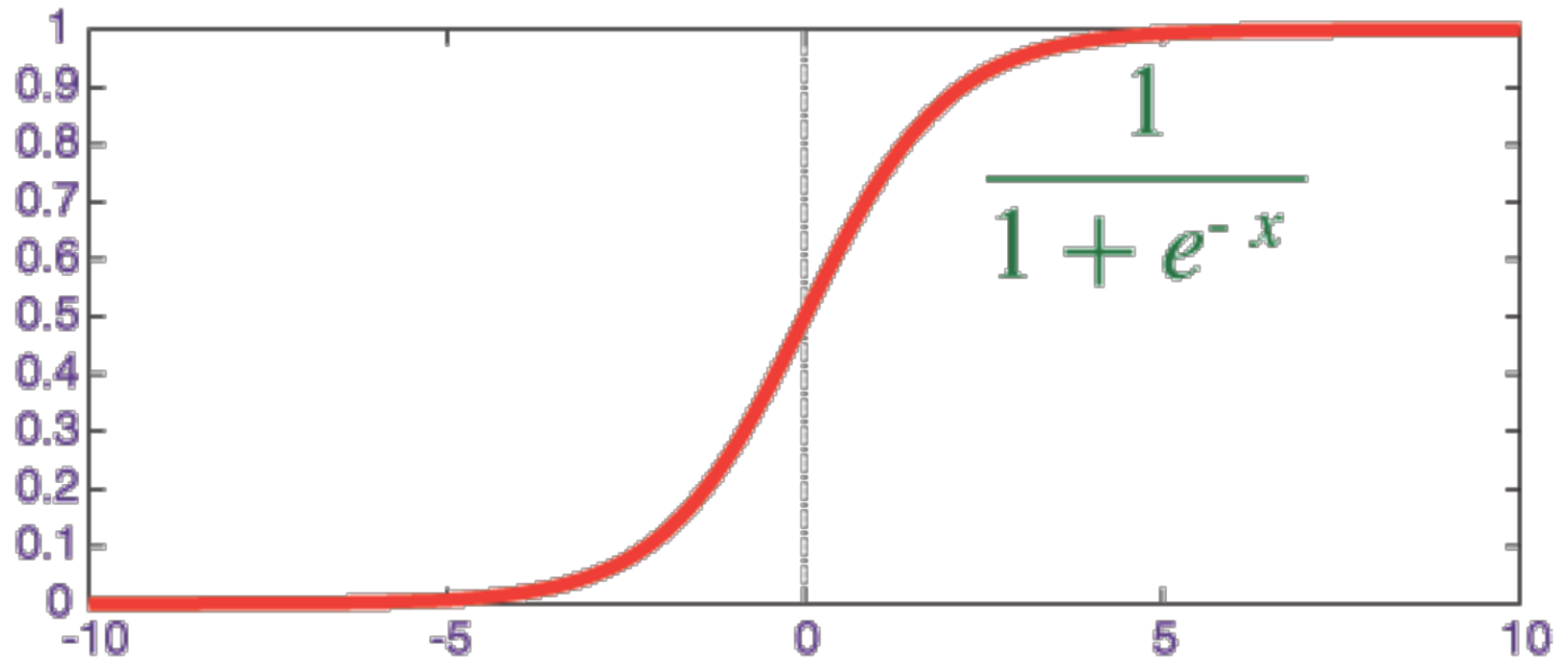
If $y_{\text{pred}} < 0.5$ predict label 0 (survived)

Logistic regression to the rescue

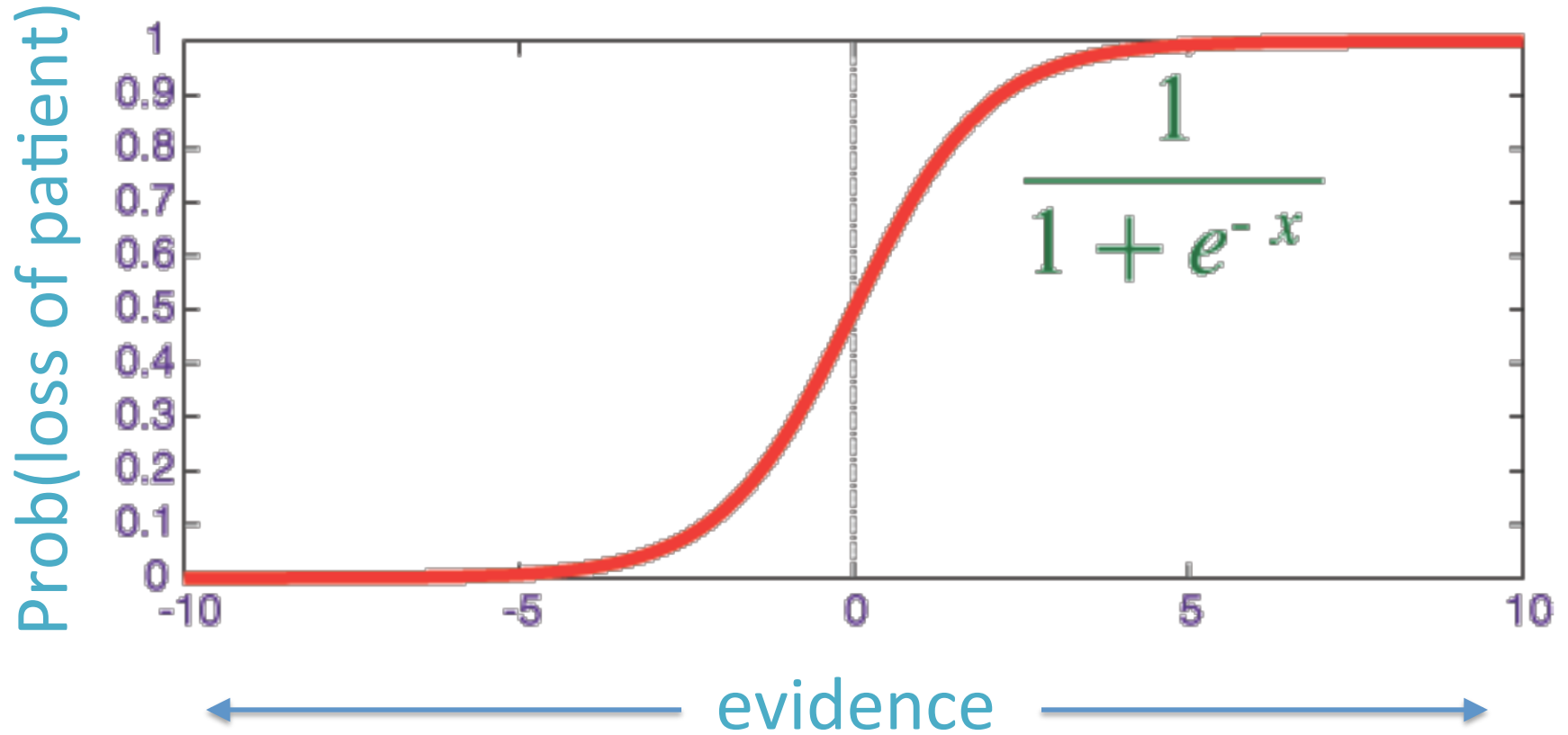


$$y_{\beta}(x) = \frac{1}{1 + e^{-(\beta_0 + \beta_1 x + \varepsilon)}}$$

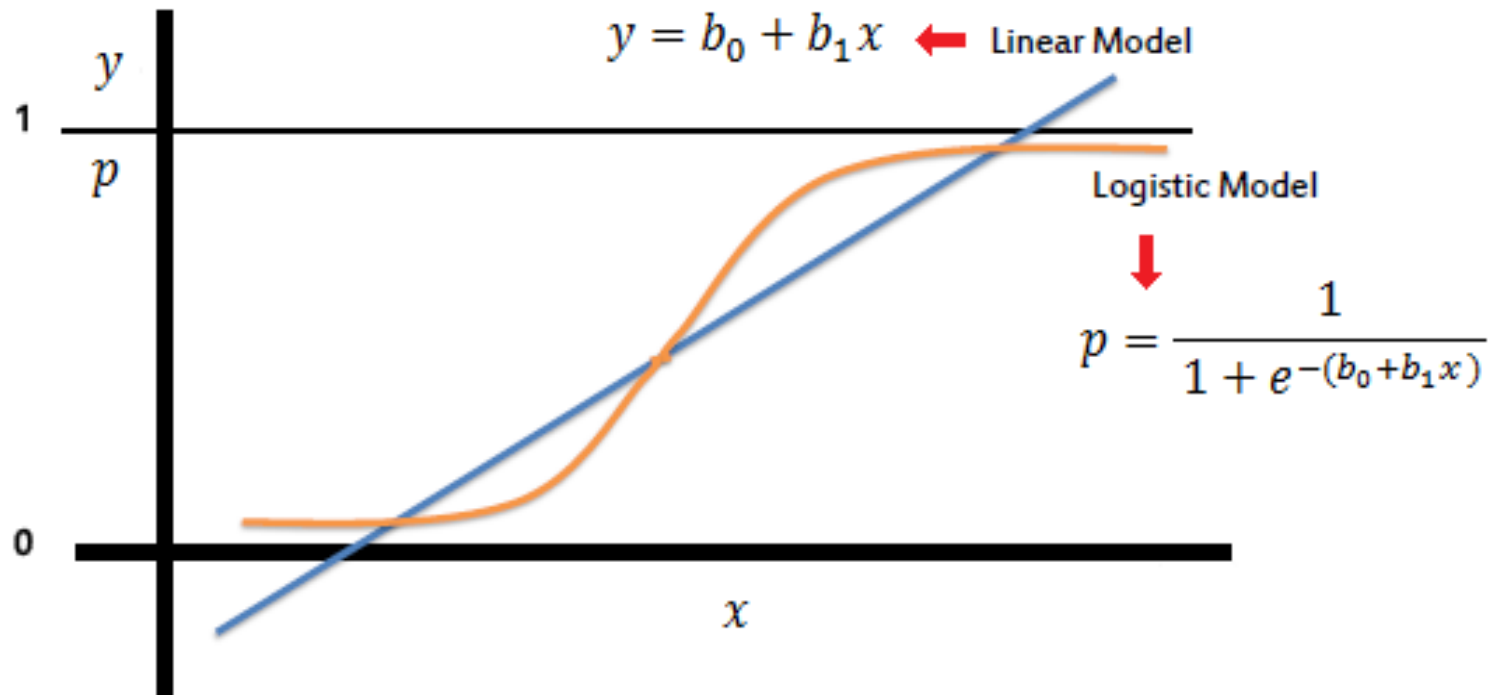
What is this function?



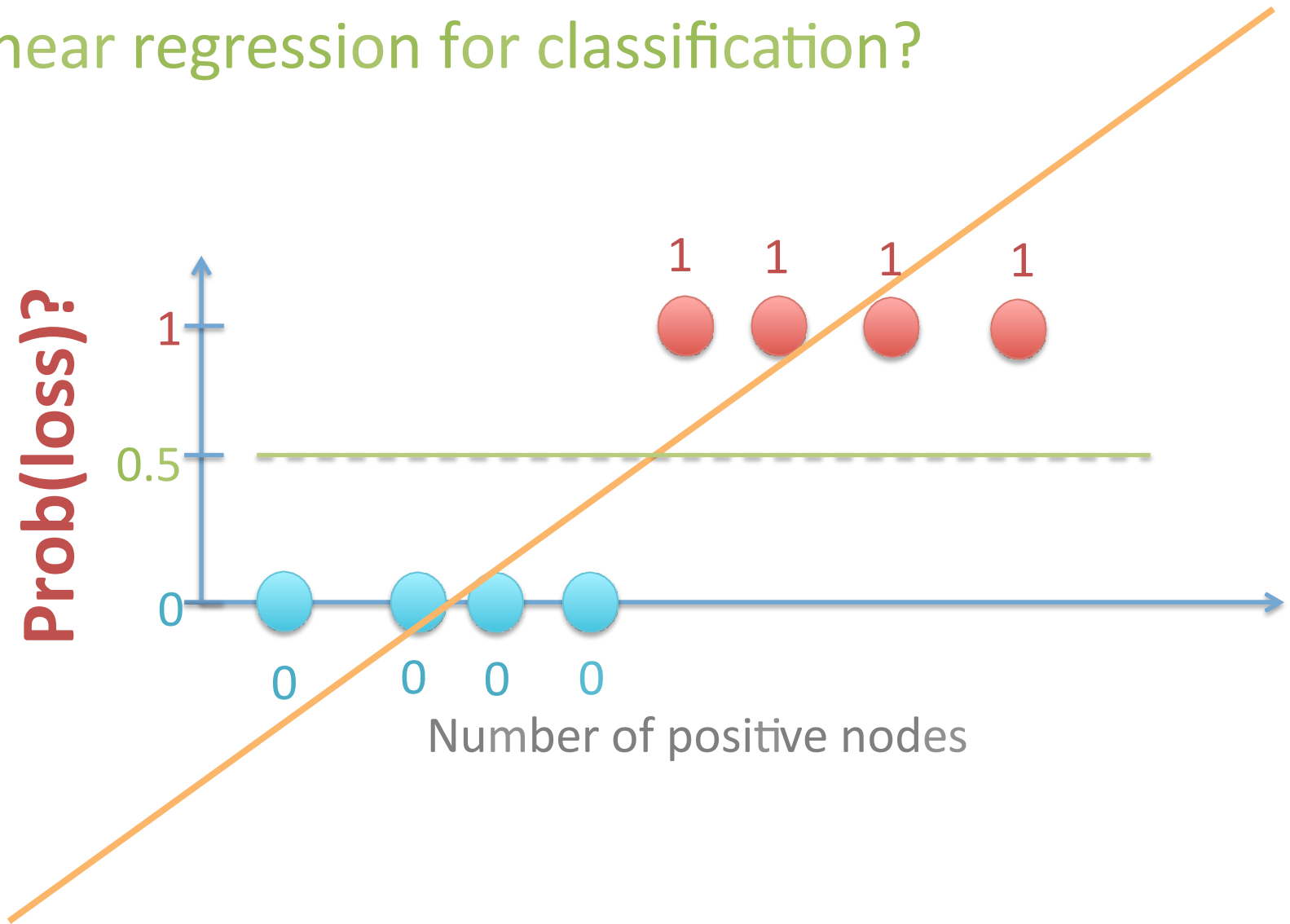
What is this function?



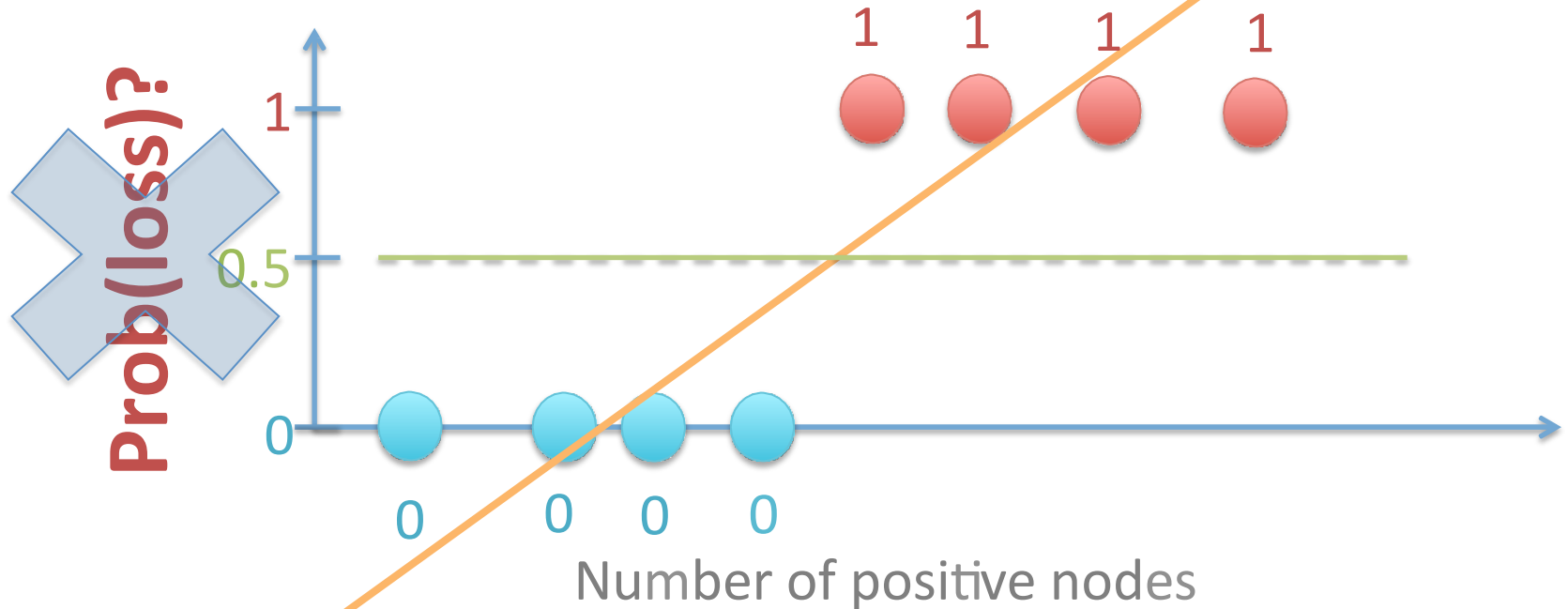
What is this function?



Linear regression for classification?



Linear regression for classification?



Y between $-\infty$ and ∞ ,
Not 0 and 1

What metric would express the chances of loss/survival, but not constrained to $[0, 1]$?

$$P(\textit{loss}) = 0.8$$

$$P(\textit{survival}) = 0.2$$

Probability

What metric would express the chances of loss/survival, but not constrained to $[0, 1]$?

$$P(loss) = 0.8$$

$$P(survival) = 0.2$$

$$\frac{P(loss)}{P(survival)} = 4$$

Probability

Odds

What metric would express the chances of loss/survival, but not constrained to $[0, 1]$?

$$\begin{array}{ll} P(loss) = 0.05 & \frac{P(loss)}{P(survival)} = 0.053 \\ P(survival) = 0.95 & \end{array}$$

Probability

Odds

What metric would express the chances of loss/survival, but not constrained to $[0, 1]$?

$$P(loss) = 0.5$$

$$P(survival) = 0.5$$

Probability

$$\frac{P(loss)}{P(survival)} = 1$$

Odds

What metric would express the chances of loss/survival, but not constrained to $[0, 1]$?

$$P(\textit{loss}) = 0.5$$

$$P(\textit{survival}) = 0.5$$

$$\frac{P(\textit{loss})}{P(\textit{survival})} = 1$$

Probability

Odds

between 0 and inf

What metric would express the chances of loss/survival, but not constrained to $[0, 1]$?

$$P(loss) = 0.5$$

$$P(survival) = 0.5$$

$$\log\left(\frac{P(loss)}{P(survival)}\right) = 0$$

Probability

Log Odds

between $-\infty$ and ∞

What metric would express the chances of loss/survival, but not constrained to $[0, 1]$?

$$\begin{aligned} P(\textit{loss}) &= 0.05 \\ P(\textit{survival}) &= 0.95 \end{aligned} \quad \log \left(\frac{P(\textit{loss})}{P(\textit{survival})} \right) = -2.94$$

Probability

Log Odds

between $-\infty$ and ∞

What metric would express the chances of loss/survival, but not constrained to $[0, 1]$?

$$P(\textit{loss}) = 0.8$$

$$P(\textit{survival}) = 0.2$$

$$\log\left(\frac{P(\textit{loss})}{P(\textit{survival})}\right) = 1.39$$

Probability

Log Odds

between $-\infty$ and ∞

What metric would express the chances of loss/survival, but not constrained to $[0, 1]$?

$$\begin{array}{l} P(\textit{loss}) = 0.999 \\ P(\textit{survival}) = 0.001 \end{array} \quad \log \left(\frac{P(\textit{loss})}{P(\textit{survival})} \right) = 6.9$$

Probability

Log Odds

between $-\infty$ and ∞

What metric would express the chances of loss/survival, but not constrained to $[0, 1]$?

$$P(loss) = 0.999$$

$$1 - P(loss) = 0.001$$

$$\log\left(\frac{P(loss)}{1 - P(loss)}\right) = 6.9$$

Probability

Log Odds

logit function

What metric would express the chances of loss/survival, but not constrained to $[0, 1]$?

$$P(loss) = 0.999$$

$$1 - P(loss) = 0.001$$

$$\log\left(\frac{P(loss)}{1 - P(loss)}\right) = 6.9$$

Probability

Log Odds

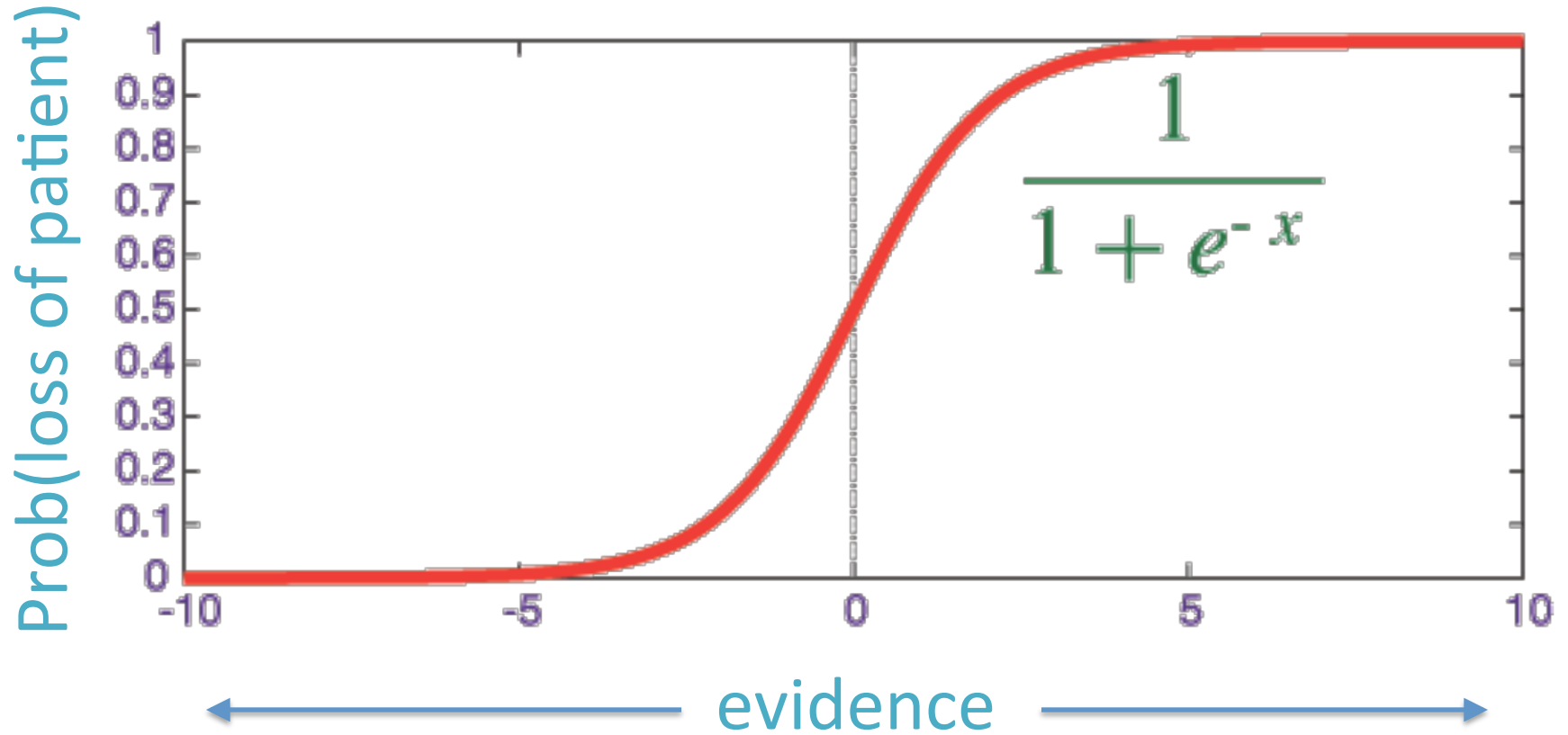
logit function

$$\frac{1}{1 + e^{-\log\left(\frac{P(loss)}{1 - P(loss)}\right)}} = P(loss)$$

Logistic Function

Log Odds \rightarrow Prob

What is this function?



“Logistic regression” is a classification algorithm

$$y_{\beta}(x) = \frac{1}{1 + e^{-(\beta_0 + \beta_1 x + \varepsilon)}}$$

```
from sklearn.linear_model import LogisticRegression  
#(just like LinearRegression)
```

```
from statsmodels.formula.api import Logit  
#(just like OLS)
```