



However, we don't know the optimal dosage to give to patients.

Subheading goes here

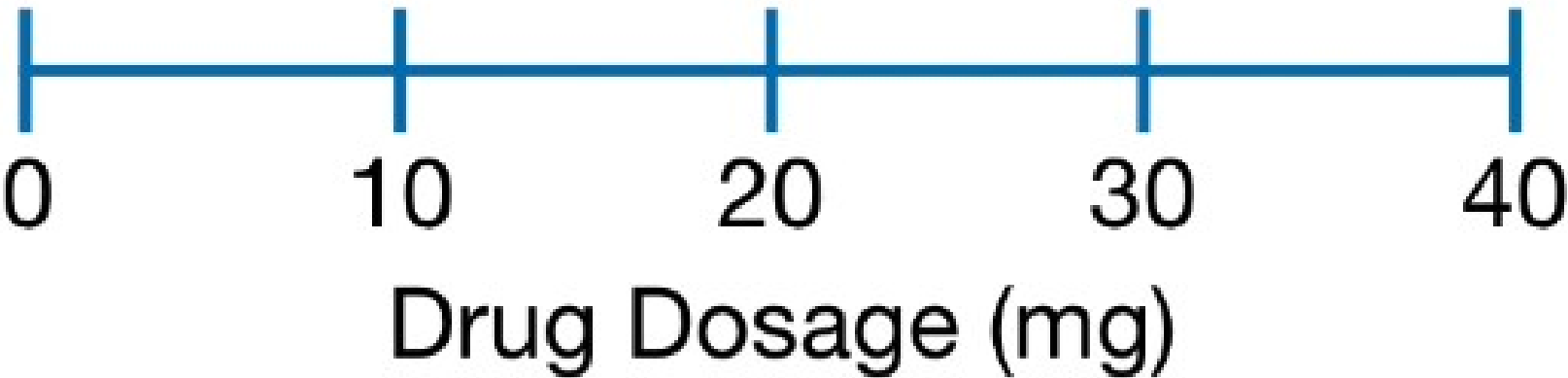


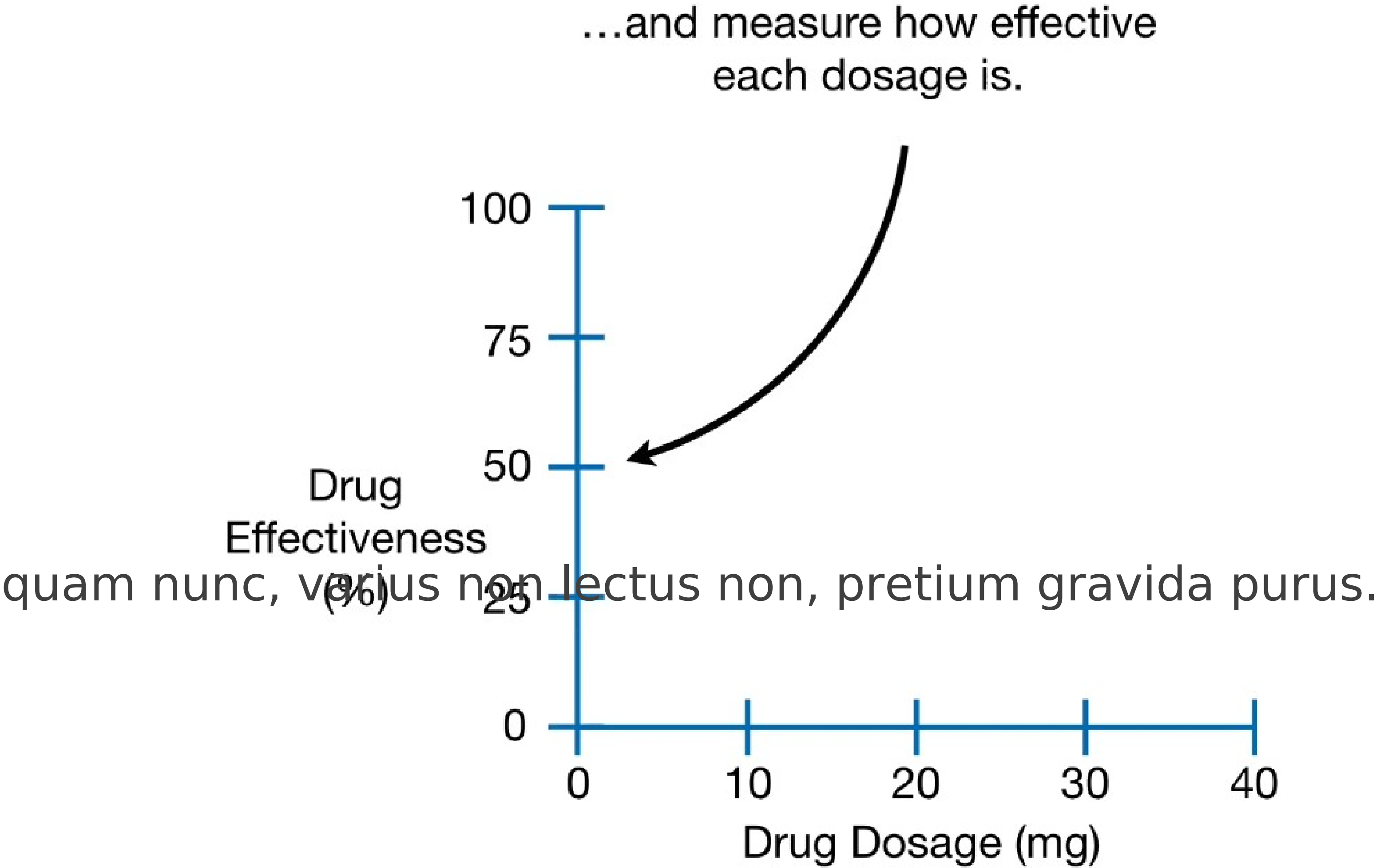


So we do a clinical trial with
different dosages...



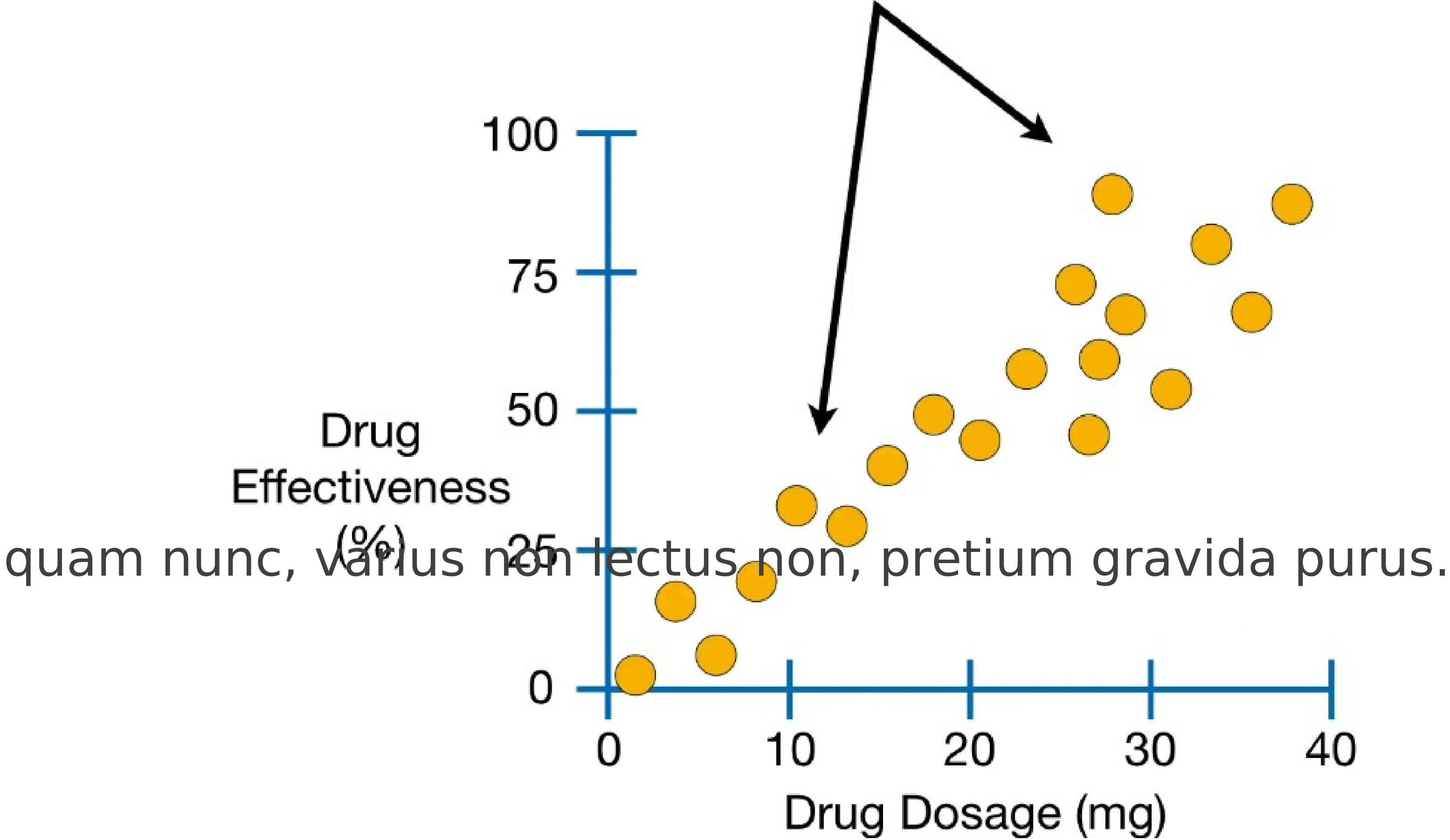
quam nunc, varius non lectus non, pretium gravida purus.





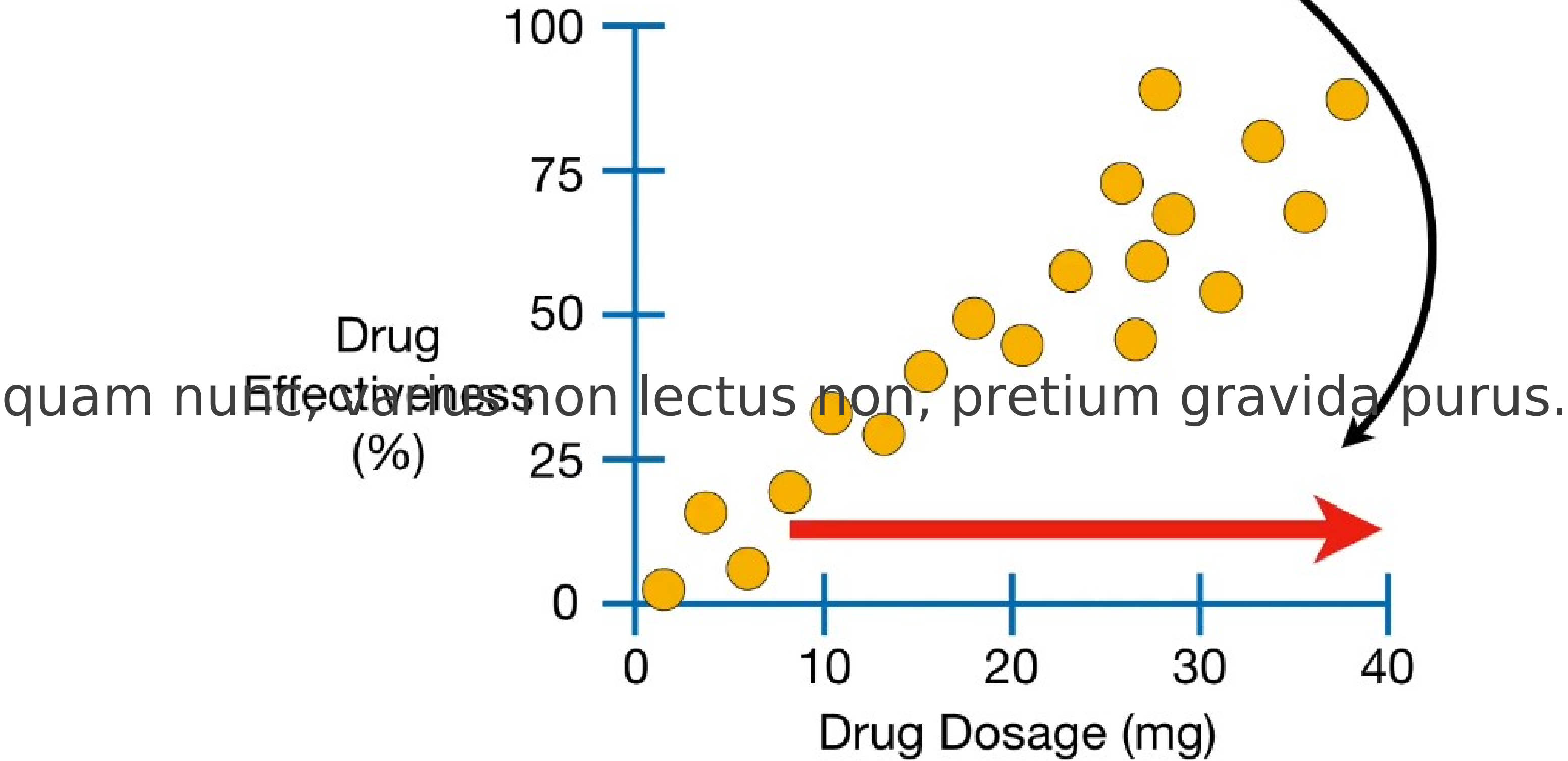


If the data looked like this...





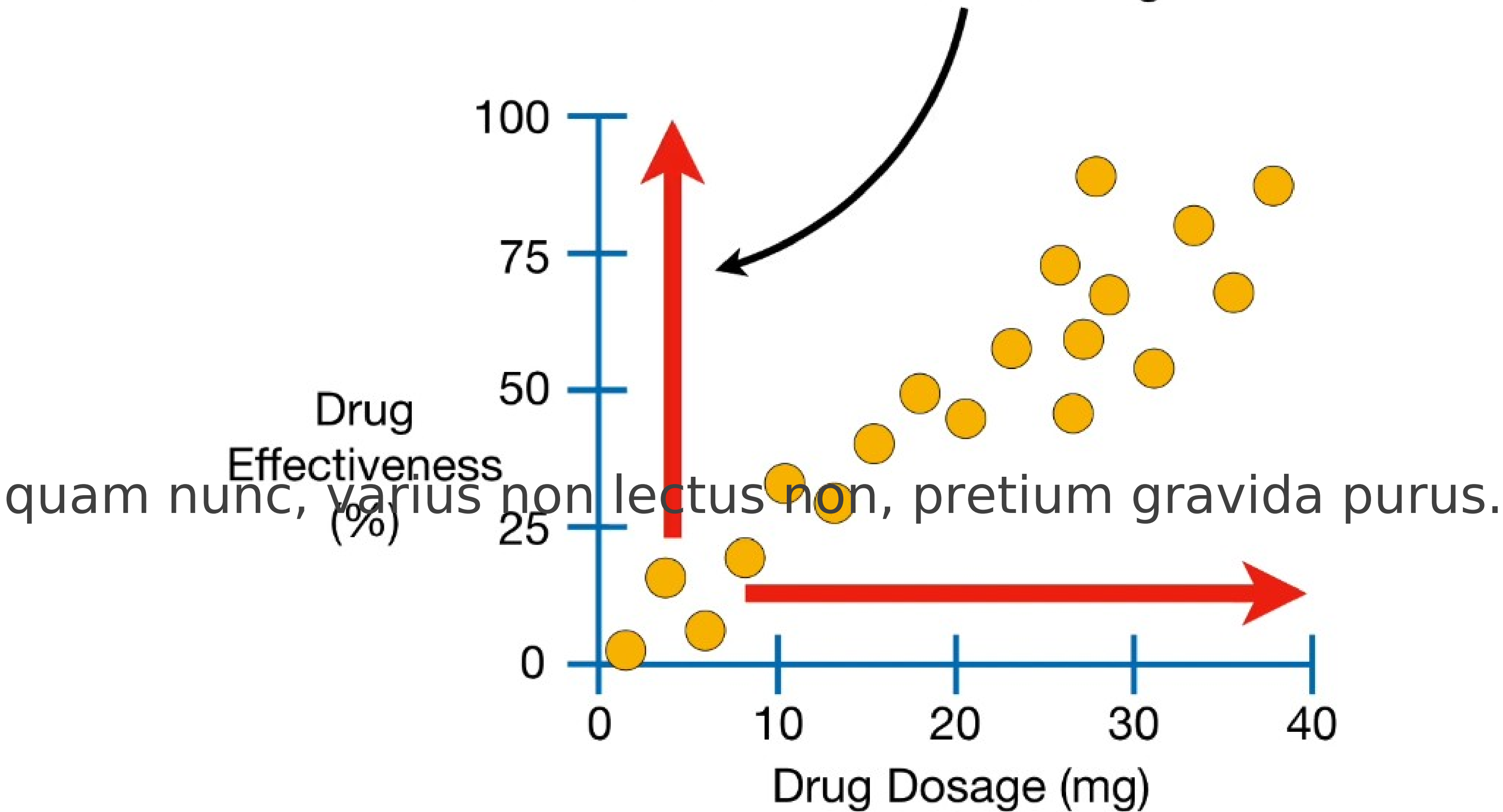
...and, in general, the higher the dose,



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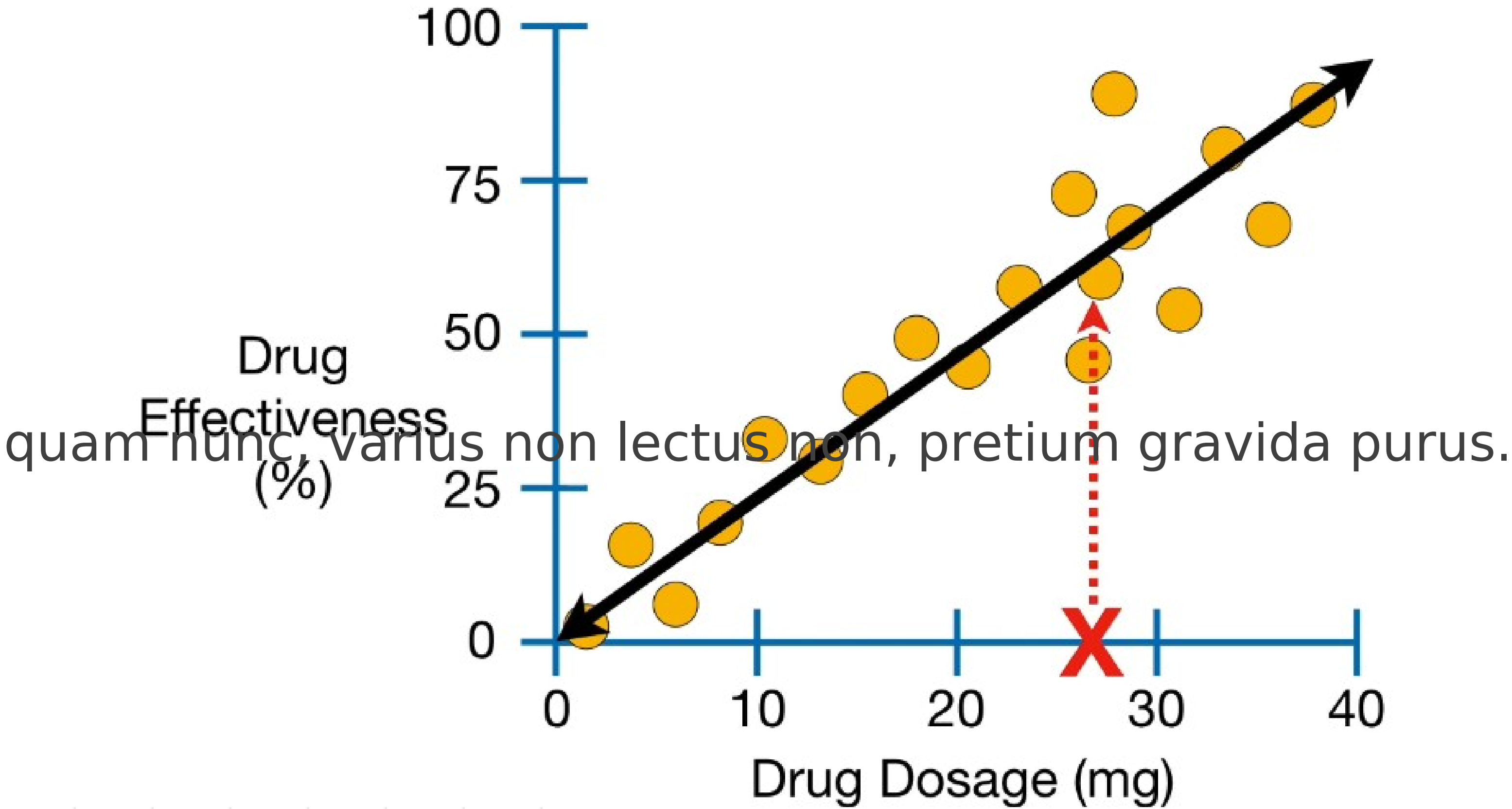
...and, in general, the higher the dose,
the more effective the drug...

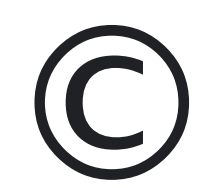


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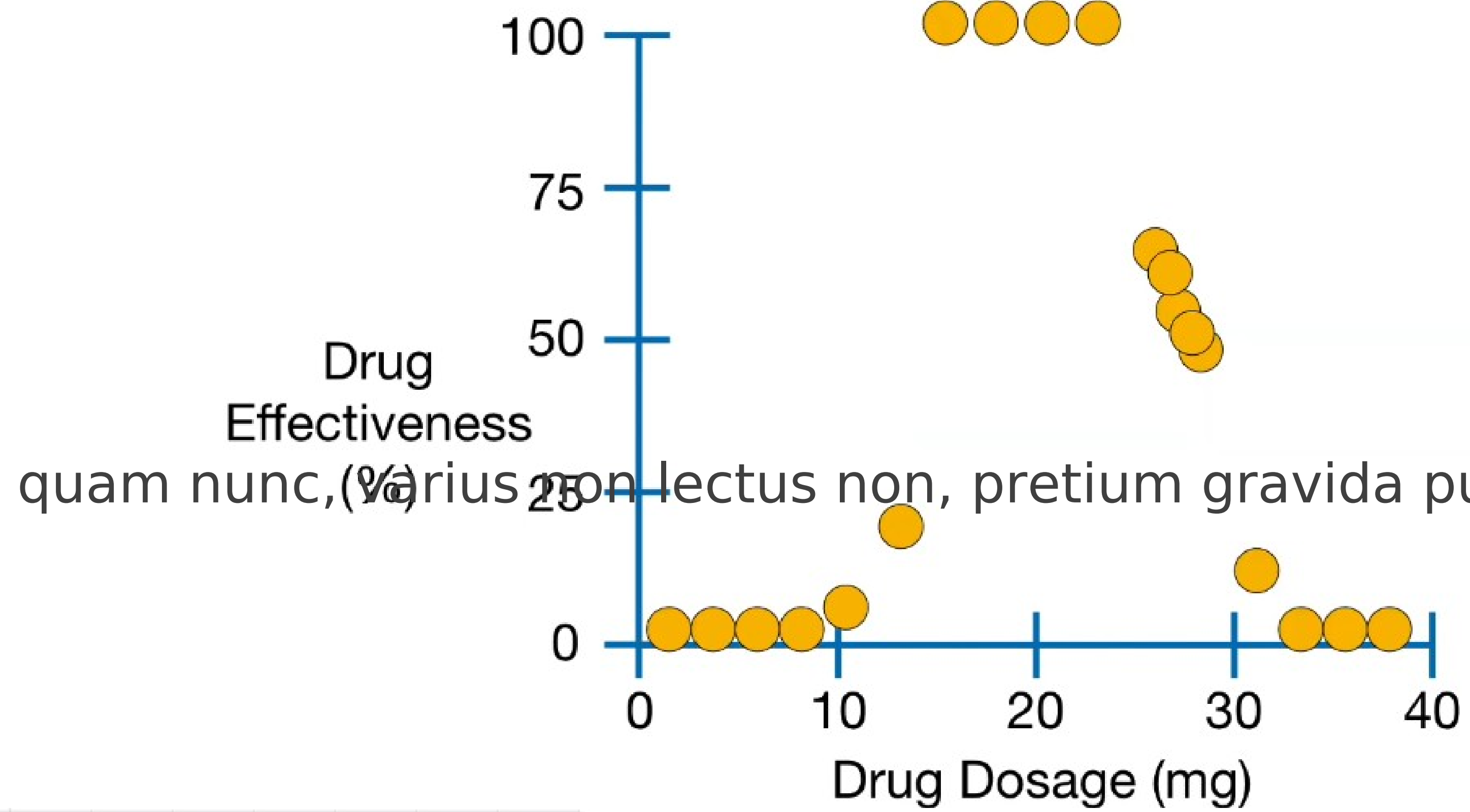


...we could use the line to predict that a **27 mg Dose** should be **62% Effective**.



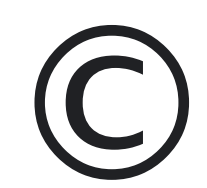


However, what if the data
looked like this?

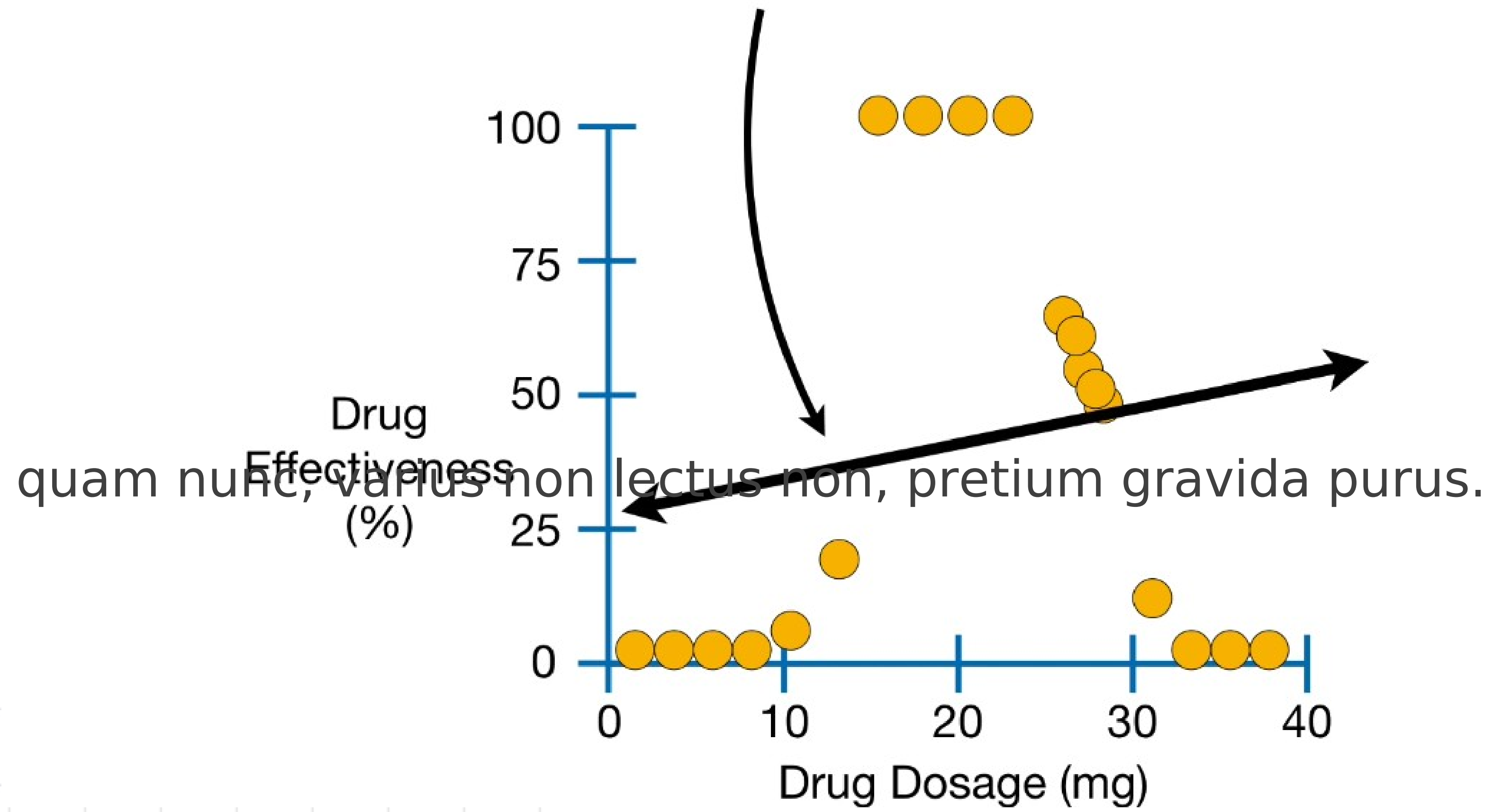


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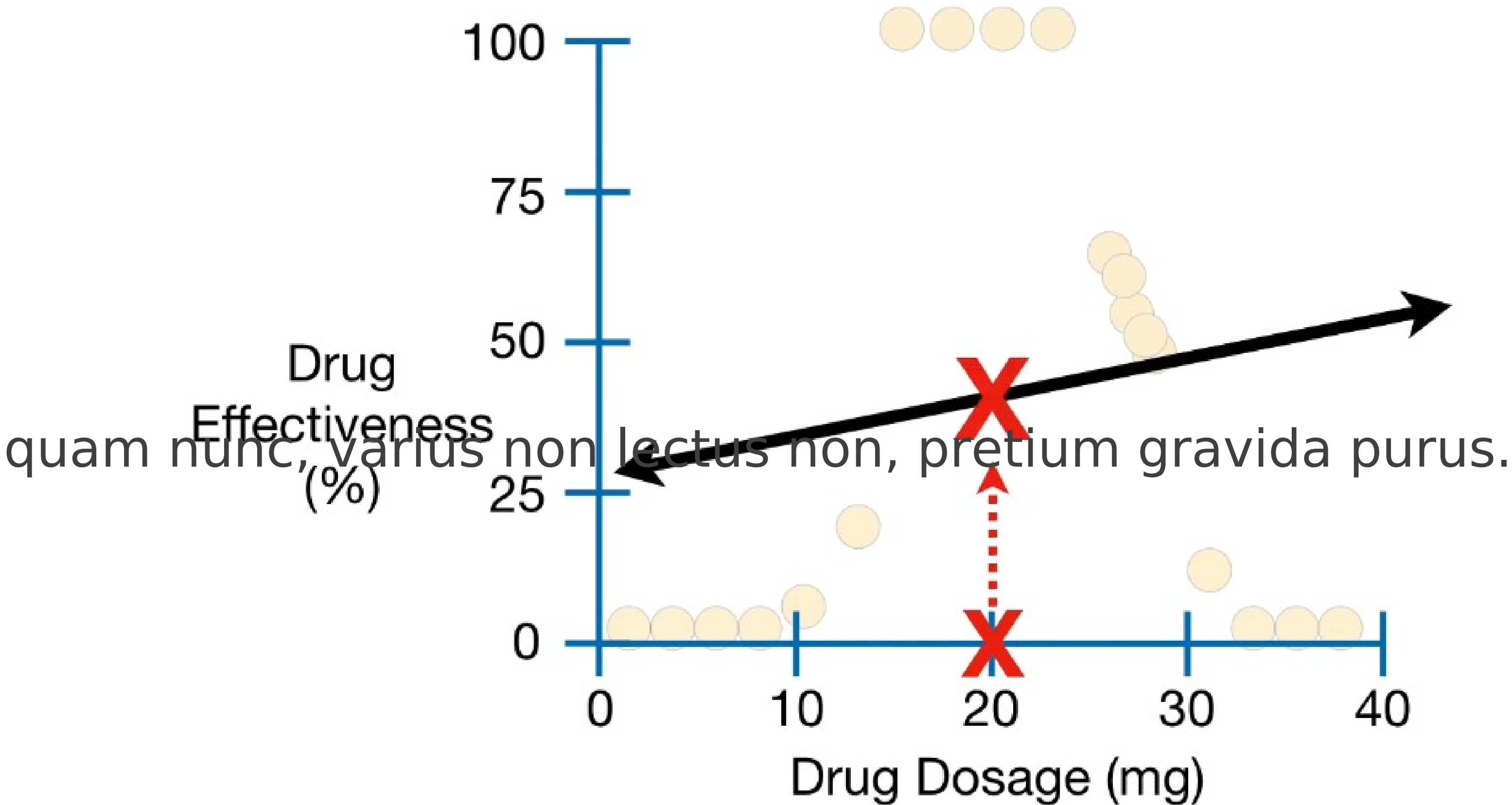
In this case, fitting a straight line to the data will not be very useful.

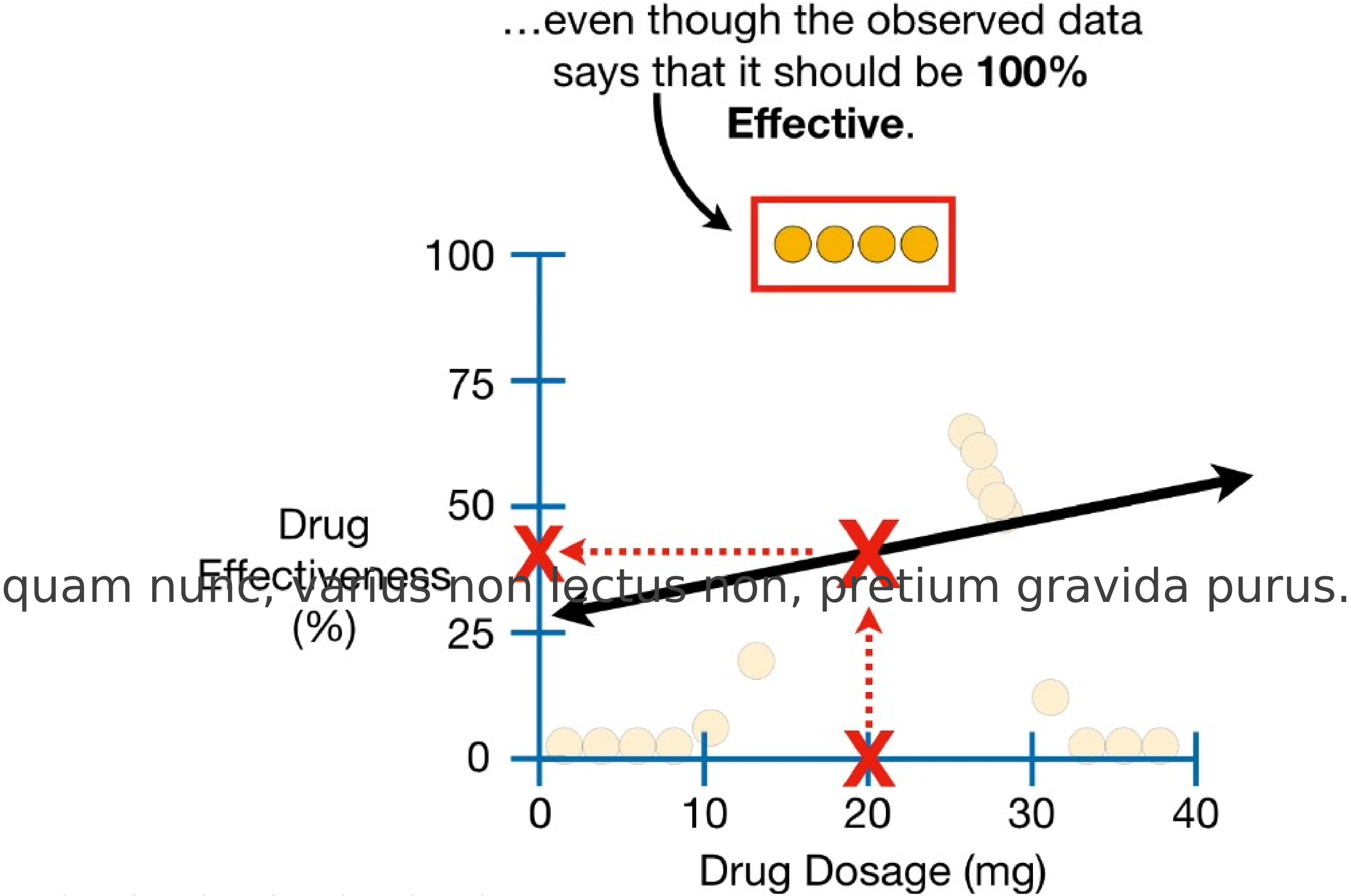


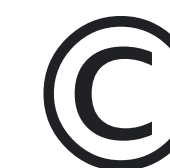
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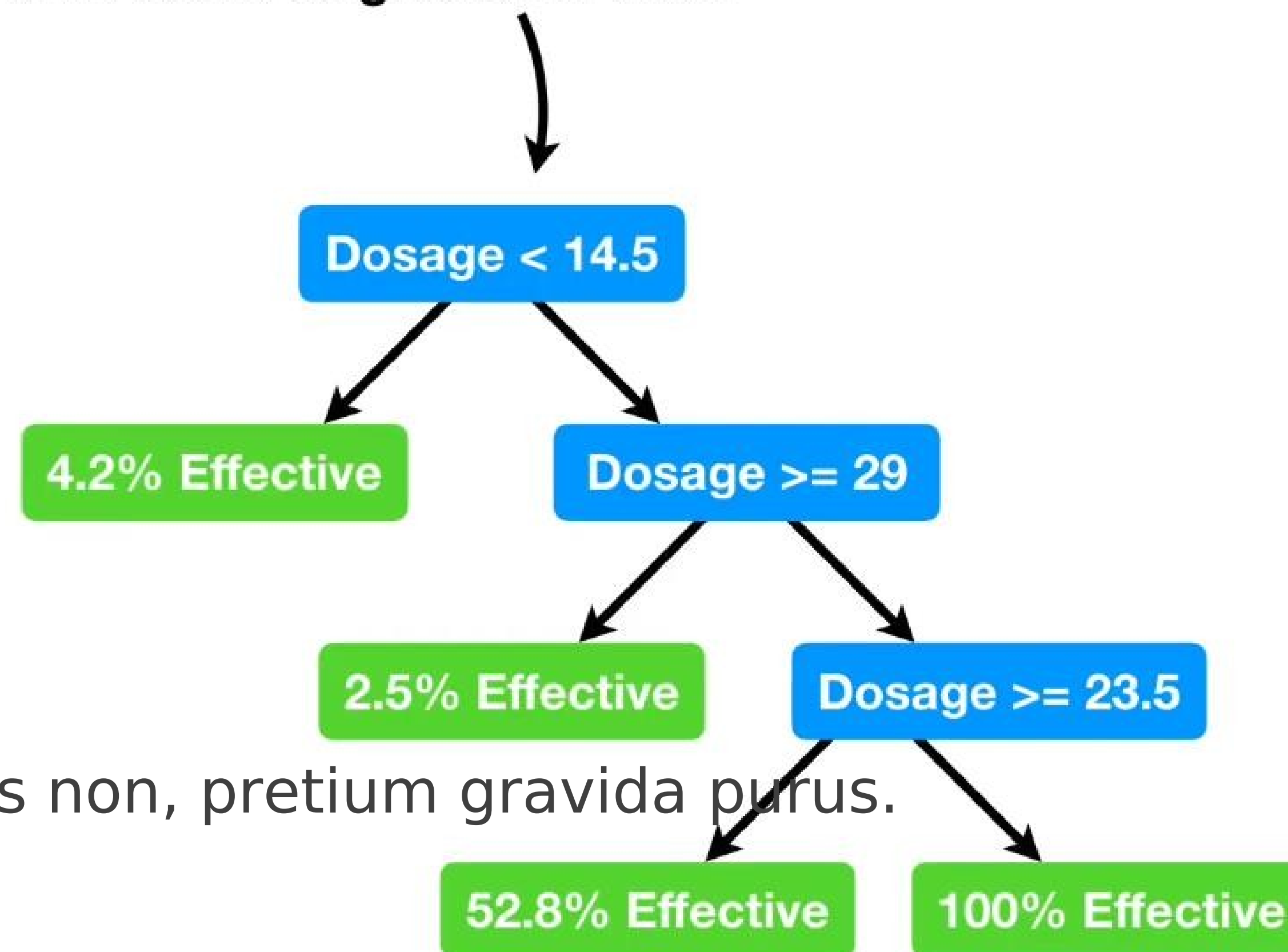
...then we would predict that a **20 mg Dose** should be **45% Effective**...







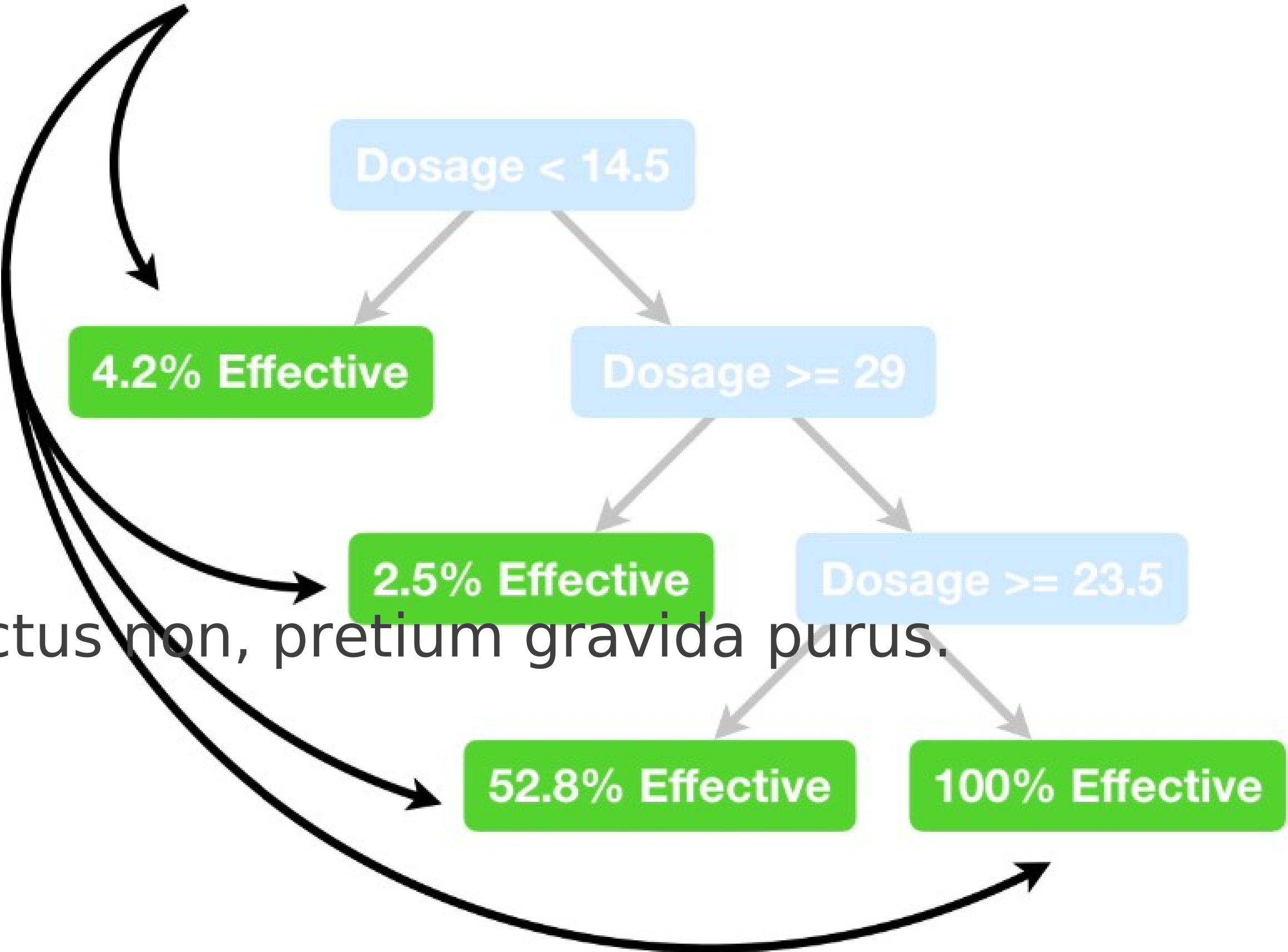
One option is to use a **Regression Tree**.



quam nunc, varius non lectus non, pretium gravida purus.



In a **Regression Tree**, each leaf represents a numeric value.

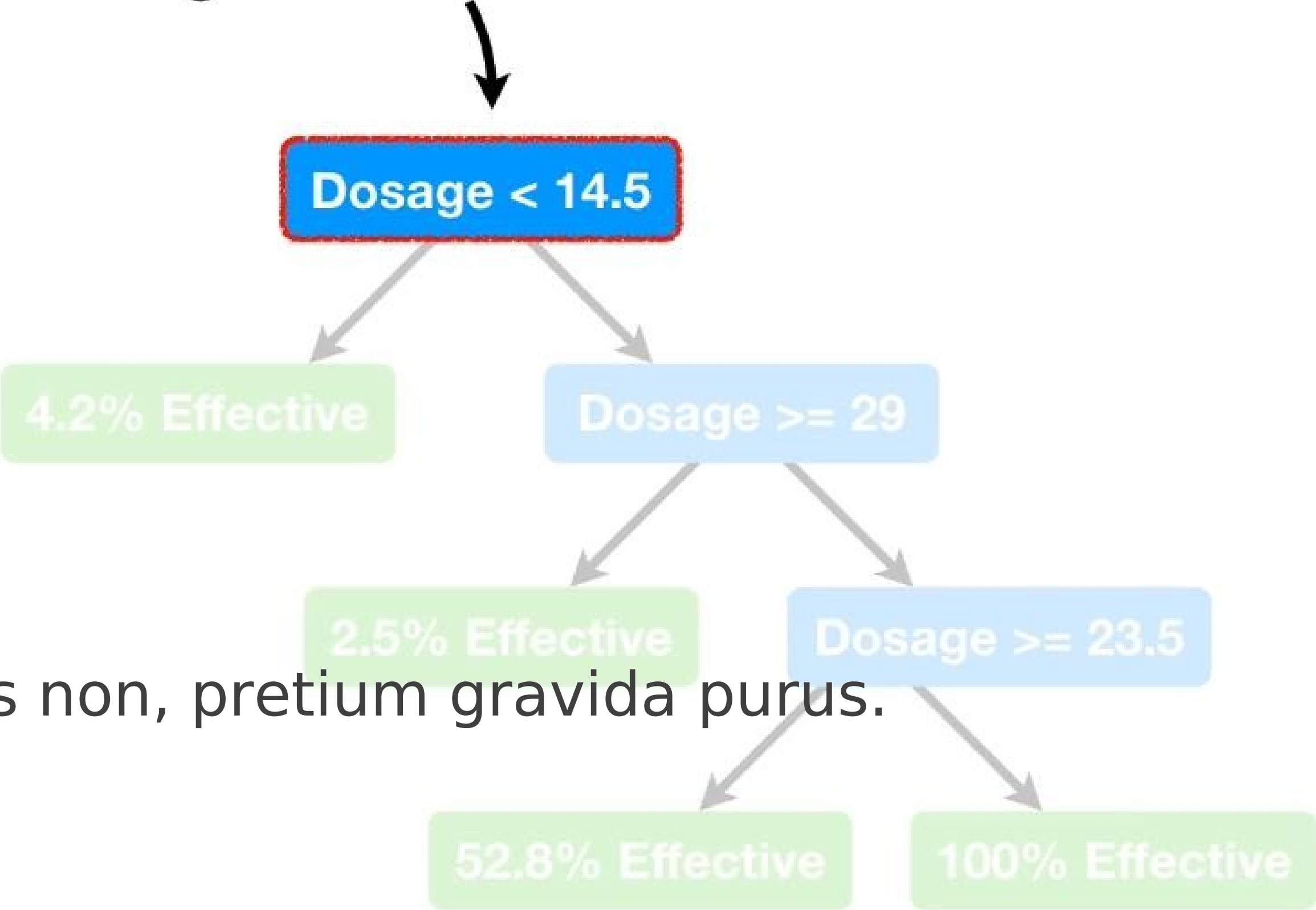


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With *this* **Regression Tree**, we start by asking if the **Dosage** is less than **14.5**.

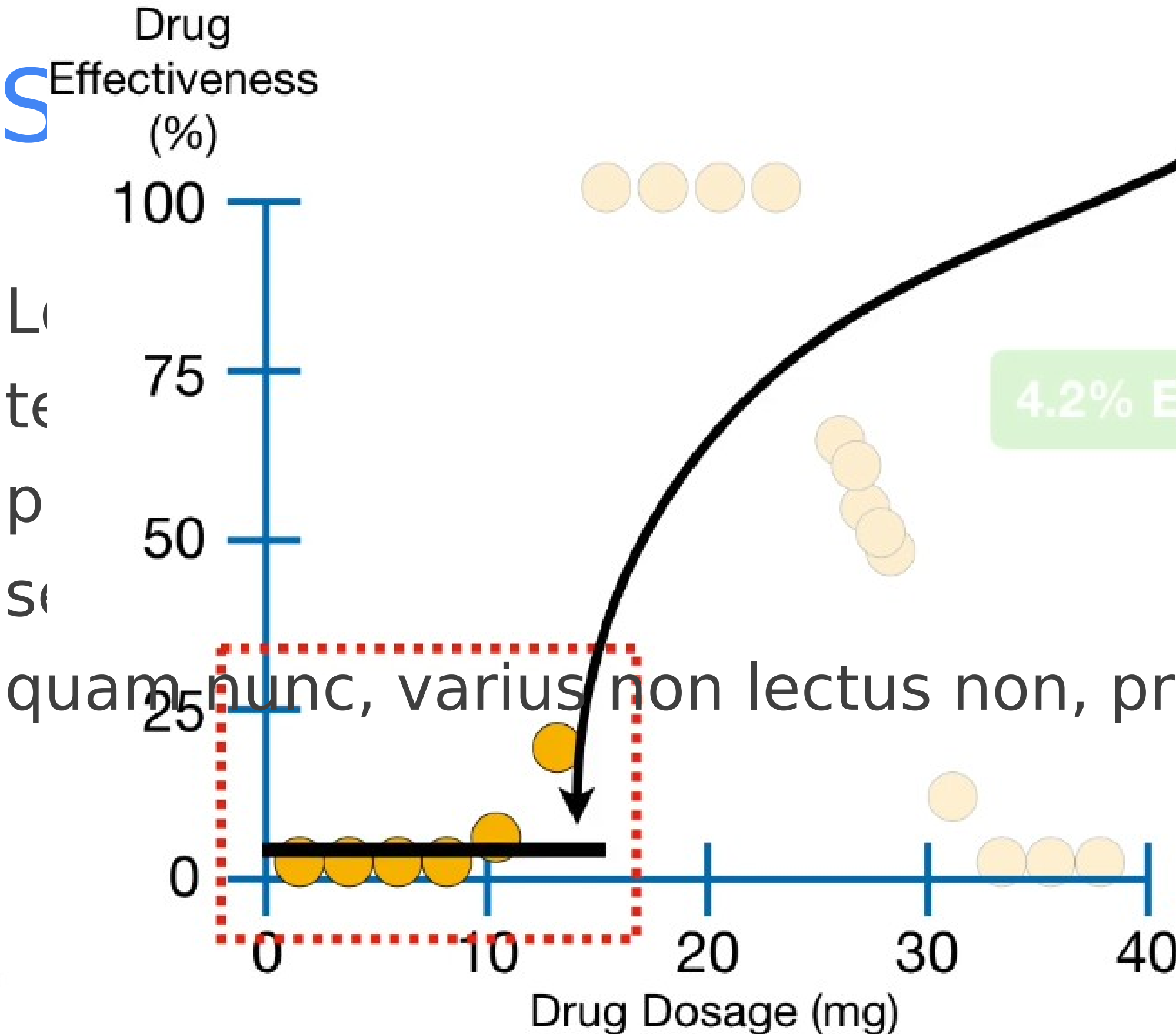


quam nunc, varius non lectus non, pretium gravida purus.



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...and the average **Drug Effectiveness** for these **6** observations is **4.2%**...



Dosage < 14.5

4.2% Effective

Dosage >= 29

2.5% Effective

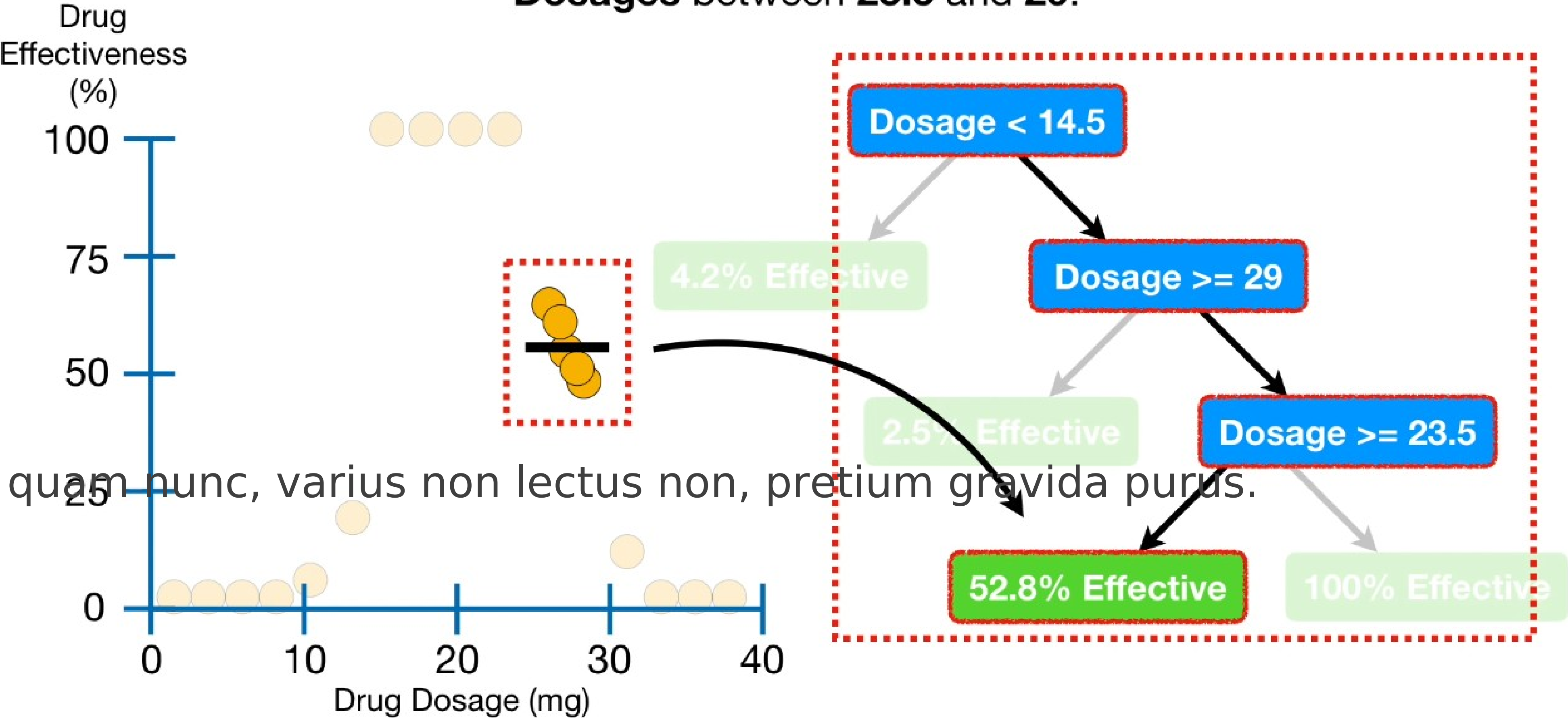
Dosage >= 23.5

52.8% Effective

100% Effective



...so the tree uses the average value,
52.8%, as its prediction for people with
Dosages between **23.5** and **29**.

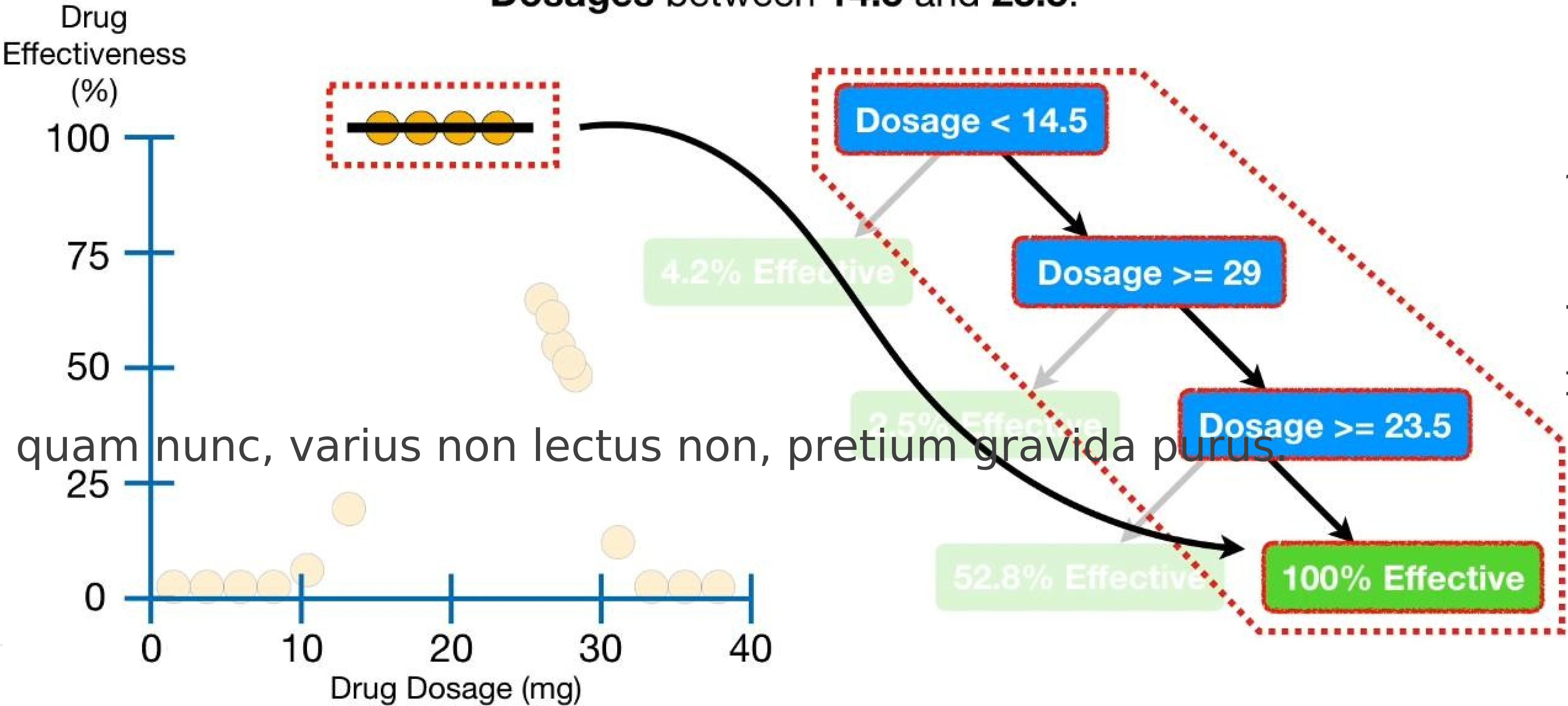


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...so the tree uses the average value,
100%, as its prediction for people with
Dosages between **14.5** and **23.5**.

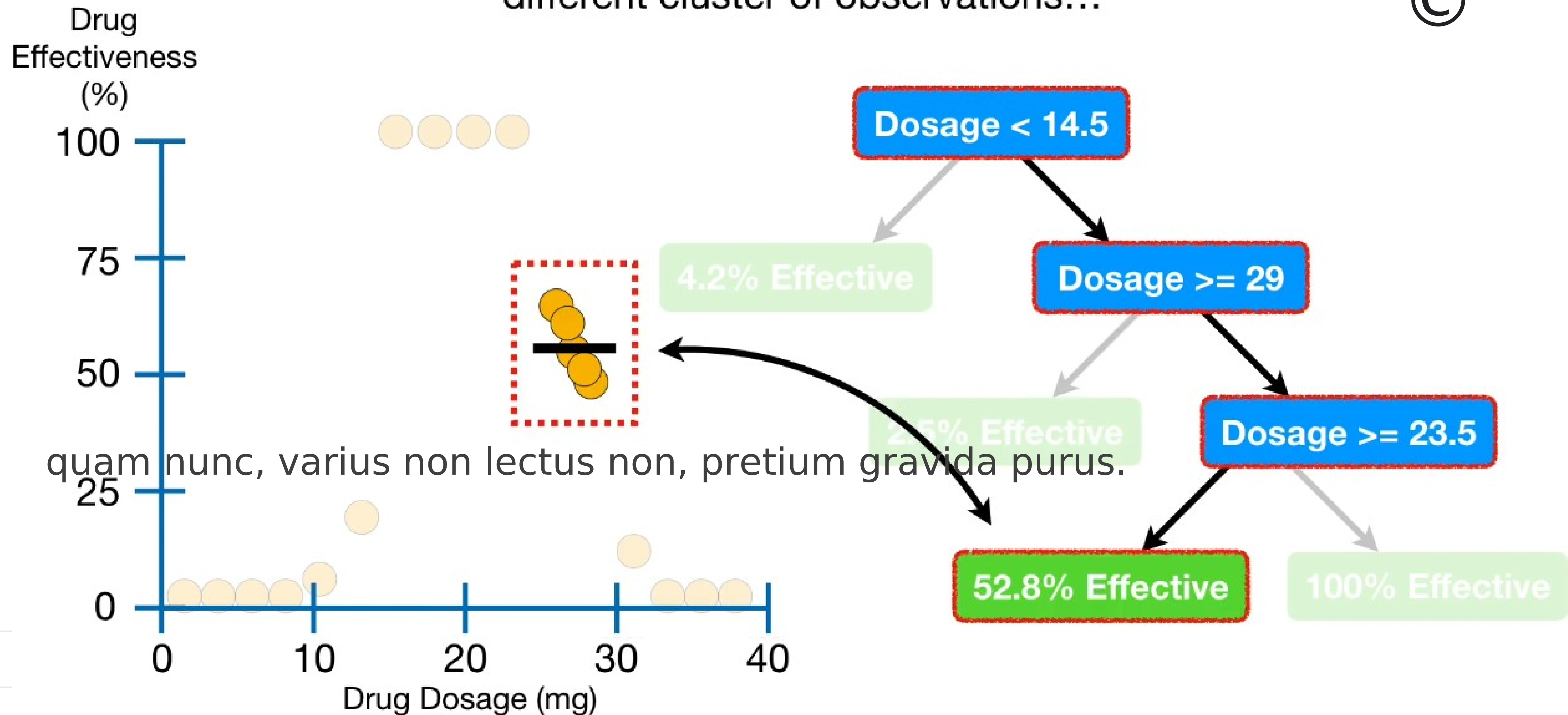


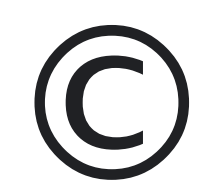
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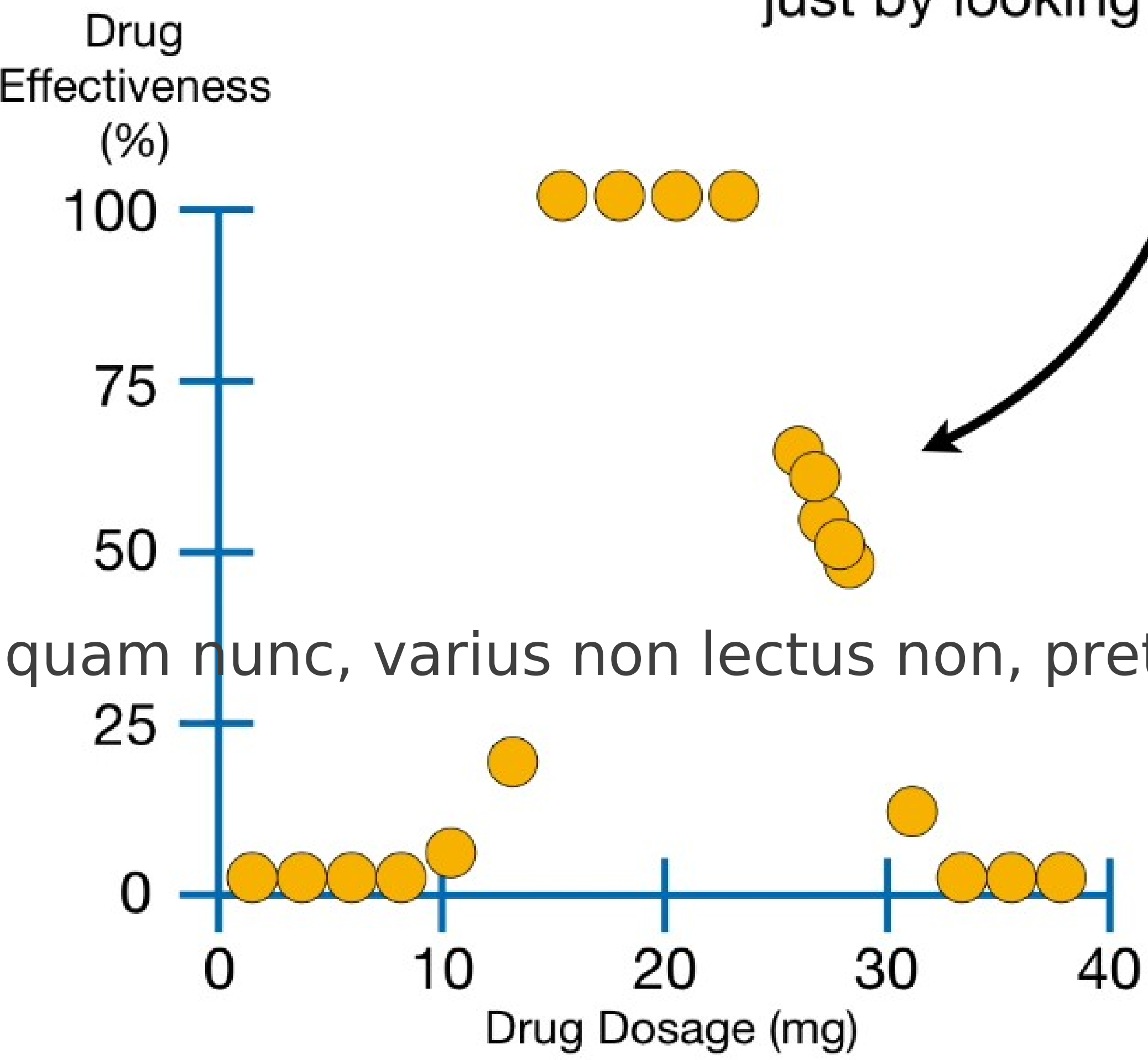
Since each leaf corresponds to the average **Drug Effectiveness** in a different cluster of observations...

Credits:
StatQuest with Josh Starmer





At this point you might be thinking, “The **Regression Tree** is cool, but I can also predict **Drug Effectiveness** just by looking at the graph...”



quam nunc, varius non lectus non, pretium gravida purus.



But when we have **3** or more predictors, like **Dosage**, **Age** and **Sex**, to predict **Drug Effectiveness**, drawing a graph is very difficult, if not impossible.

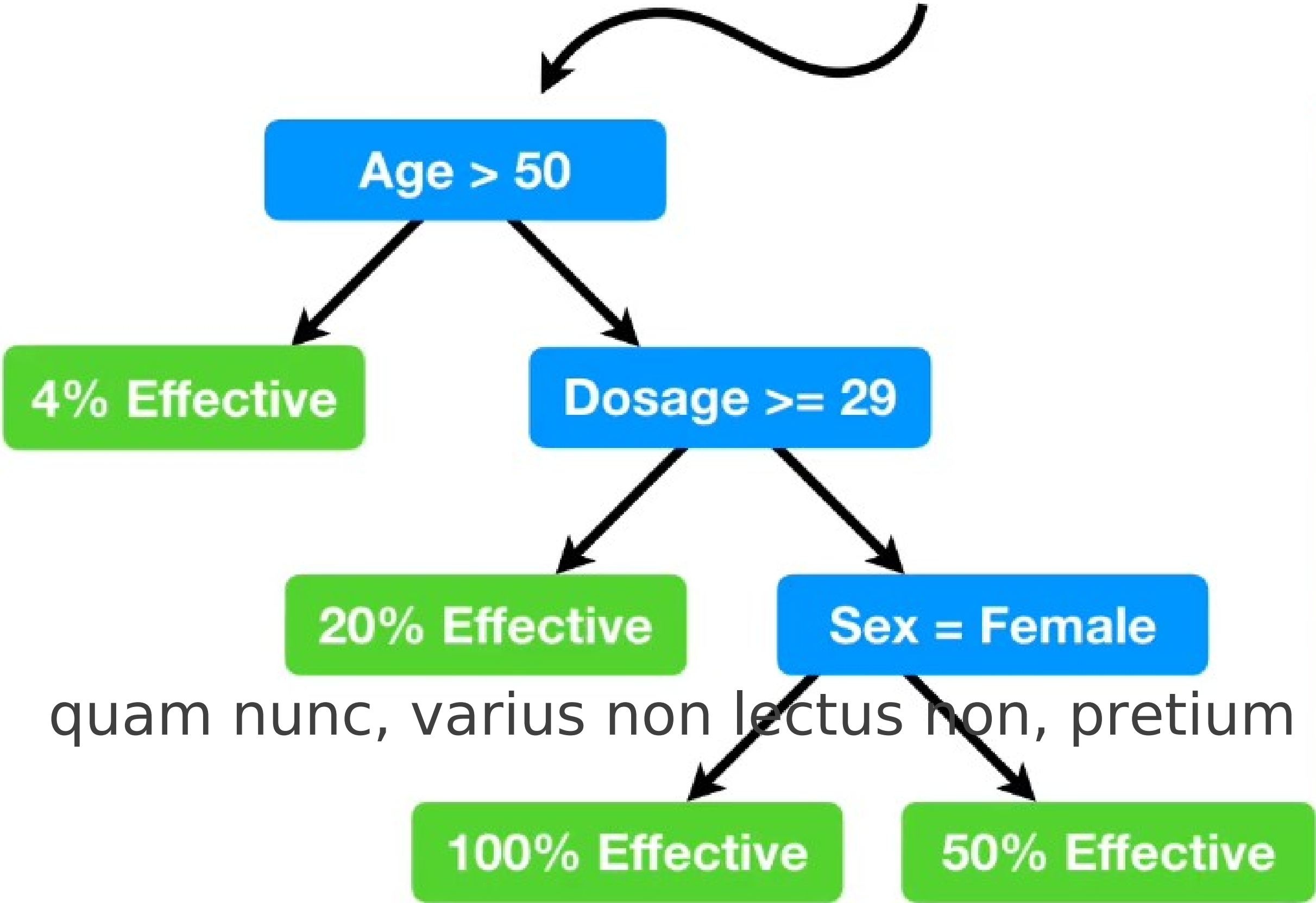


Dosage	Age	Sex	Etc.	Drug Effect.
10	25	Female	...	98
20	73	Male	...	0
35	34	Female	...	100
5	12	Male	...	44
etc...	etc...	etc...	etc...	etc...

quam nunc, varius non lectus non, pretium gravida purus

In contrast, a **Regression Tree** easily accommodates the additional predictors.

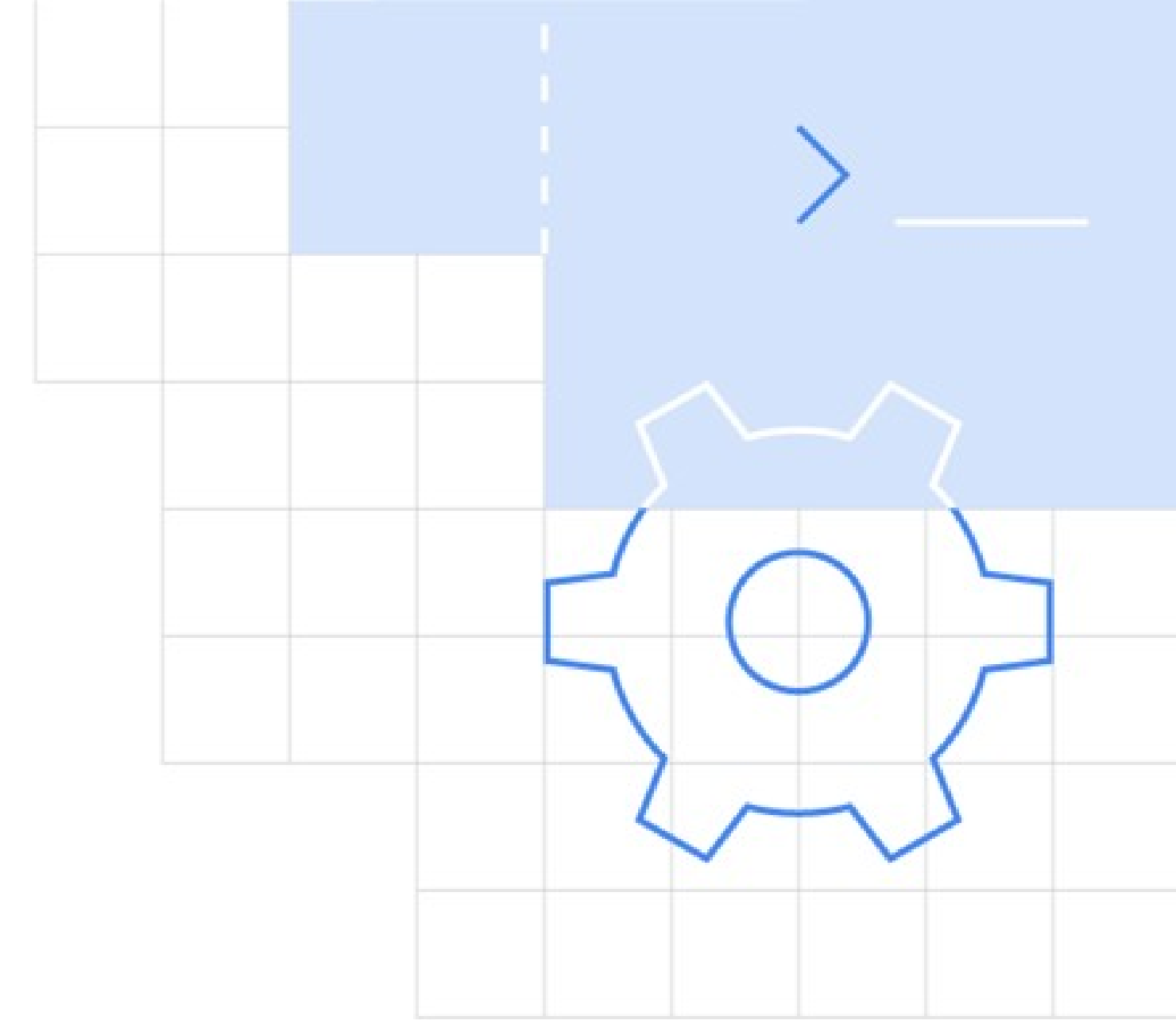
Credits:
StatQuest with Josh Starmer



quam nunc, varius non lectus non, pretium gravida purus.

Dosage	Age	Sex	Etc.	Drug Effect.
10	25	Female	...	98
20	73	Male	...	0
35	54	Female	...	100
5	12	Male	...	44
etc...	etc...	etc...	etc...	etc...

Now you fully understand
the concept



Decision Trees

Random Forest