

Topic 4  
Correlation and Regression  
Transformed Variables

# Outline

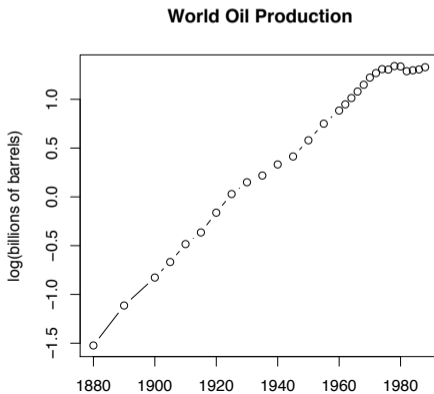
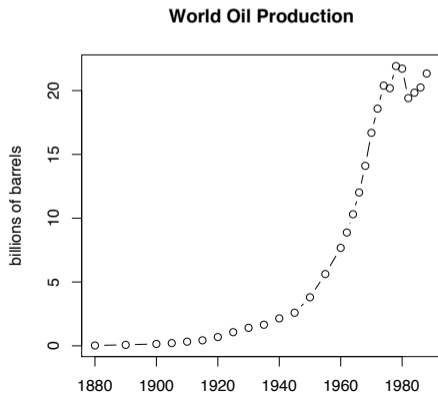
Worldwide Oil Production

Michaelis-Menten Kinetics

Lineweaver-Burke double reciprocal plot

## Worldwide Oil Production

**Example.** The modern history of petroleum began in the 19th century with the refining of kerosene from crude oil. The industry grew through the 1800s, driven by the demand for kerosene and oil lamps. The internal combustion engine introduced in the early part of the 20th century provided a demand that continues to sustained the industry.



## Worldwide Oil Production

The relationship between the **explanatory variable** and **response variable** can be made to be linear with a simple transformation, the common **logarithm**. The explanatory variable remains **year**. With these variables, we can use a regression line to help describe the data.

$$\log y_i = \alpha + \beta x_i + \epsilon_j.$$

```
> summary(lm(logbarrel~year))  
Call:  
lm(formula = logbarrel ~ year)  
Coefficients:  
                Estimate Std. Error t value Pr(>|t|)  
(Intercept) -5.159e+01  1.301e+00  -39.64  <2e-16 ***  
year          2.675e-02  6.678e-04   40.05  <2e-16 ***  
---  
Multiple R-Squared:  0.9834, Adjusted R-squared:  0.9828
```

## Worldwide Oil Production

The R output gives  $r^2 = 0.9828$ . Thus, the correlation,  $r = 0.9914$ , is very nearly one and the data lies very close to the regression line

$$\widehat{\log(\text{barrel})} = -51.59 + 0.02675 \cdot \text{year}.$$

If we rewrite the equation in exponential form, we obtain

$$\widehat{\text{barrel}} = A10^{0.02675 \cdot \text{year}} = Ae^{\hat{k} \cdot \text{year}}.$$

Thus,  $\hat{k}$  gives the instantaneous growth rate that best fits the data. This is obtained by converting from a **common logarithm** to a **natural logarithm**.

$$\hat{k} = 0.02675 \ln 10 = 0.0616$$

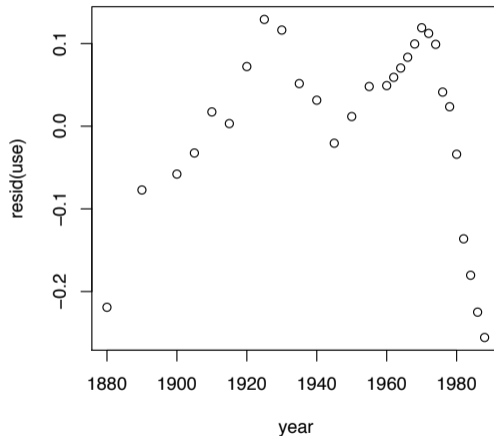
. Thus, the use of oil sustained annual growth of **6%** over a span of a hundred years.

## Worldwide Oil Production

For the residual plot.

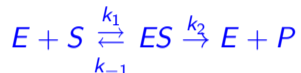
```
> use<-lm(logbarrel~year)
> plot(year,resid(use))
```

**Exercise.** Give some aspects of world history that could explain the structure in the residual plots.



## Michaelis-Menten Kinetics

Consider the chemical reaction in which an **enzyme** catalyzes the action on a **substrate**.



- $E_0$  is the **total amount of enzyme**.
- $E$  is the **free enzyme**.
- $S$  is the **substrate**.
- $ES$  is the **substrate-bound enzyme**.
- $P$  is the **product**.
- $V = d[P]/dt$  is the **production rate**.

The symbol  $[ \cdot ]$  indicates **concentration**.

## Michaelis-Menten Kinetics

The enzyme,  $E_0$ , is either free or bound to the substrate. Its **total concentration**

$$[E_0] = [E] + [ES], \quad \text{and, thus} \quad [E] = [E_0] - [ES].$$

Our goal is to relate the **production rate**  $V$  to the **substrate concentration**  $[S]$ .

The **law of mass action** turns the **chemical reactions** into **differential equations**. In particular, the reactions, focusing on the substrate-bound enzyme and the product, gives the equations

$$\frac{d[ES]}{dt} = k_1[E][S] - (k_{-1} + k_2)[ES] \quad \text{and} \quad V = \frac{d[P]}{dt} = k_2[ES].$$

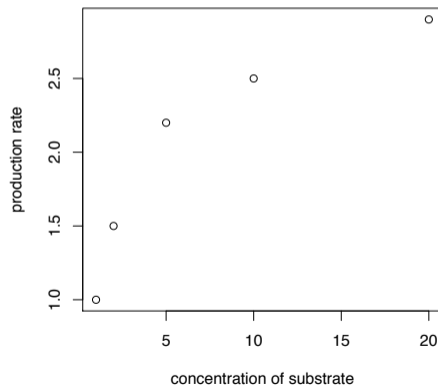
We can meet our goal by finding an equation for  $V = k_2[ES]$  that depends only on  $[S]$ .



## Michaelis-Menten Kinetics

Let's look at data,

$[S]$ (mM)	1	2	5	10	20
$V$ (nmol/sec)	1.0	1.5	2.2	2.5	2.9



## Michaelis-Menten Kinetics

To use linear regression, we will have to transform the data. The **Michaelis-Menten transformation** applies to situations in which the concentration of the substrate-bound enzyme changes much more slowly than those of the product and substrate.

$$0 \approx \frac{d[ES]}{dt} = k_1[E][S] - (k_{-1} + k_2)[ES]$$

$$[ES] \approx \frac{k_1[E][S]}{k_{-1} + k_2} = \frac{[E][S]}{K_m} = \frac{([E_0] - [ES])[S]}{K_m},$$

The ratio  $K_m = (k_{-1} + k_2)/k_1$  of the rate of loss of the substrate-bound enzyme to its production is called the **Michaelis constant**.

## Michaelis-Menten Kinetics

We now solve for  $[ES]$ .

$$[ES] \approx \frac{([E_0] - [ES])[S]}{K_m} \quad \text{and} \quad [ES] \approx [E_0] \frac{[S]}{K_m + [S]}$$

Under this approximation, the production rate of the product is:

$$V = \frac{d[P]}{dt} = k_2[ES] = k_2[E_0] \frac{[S]}{K_m + [S]} = V_{\max} \frac{[S]}{K_m + [S]}$$

Here,  $V_{\max} = k_2[E_0]$  is the **maximum production rate**. To perform linear regression, we need to have a function of  $V$  be linearly related to a function of  $[S]$ . This is achieved via taking the **reciprocal** of both sides of this equation.

$$\frac{1}{V} = \frac{[S] + K_m}{V_{\max}[S]} = \frac{1}{V_{\max}} + \frac{K_m}{V_{\max}} \frac{1}{[S]}$$

## Lineweaver-Burke double reciprocal plot

Thus, we have a *linear relationship* between

$\frac{1}{V}$ , the *response variable*, and  $\frac{1}{[S]}$ , the *explanatory variable*

subject to experimental error. This linear relationship, called the **Lineweaver-Burke double reciprocal plot**, provides a useful method for analysis of the Michaelis-Menten equation. For the data,

```
> S<-c(1,2,5,10,20);V<-c(1.0,1.5,2.2,2.5,2.9)
> Sinv<-1/S; Vinv<-1/V
> lm(Vinv~Sinv)
```

Call:

```
lm(formula = Vinv ~ Sinv)
```

Coefficients:

(Intercept)	Sinv
0.3211	0.6813



## Exercises

**Exercise.** Use the regression line  $\widehat{\log(\textit{barrel})} = -51.59 + 0.02675 \cdot \textit{year}$  to predict use of oil today. Compare it to actual use.

**Exercise.** Give the Lineweaver-Burke double reciprocal plot for

$$\frac{\widehat{1}}{V} = \frac{1}{V_{\max}} + \frac{K_m}{V_{\max}} \frac{1}{[S]} = 0.3211 + 0.6813 \cdot \frac{1}{[S]}.$$

Determine  $K_m$  and  $V_{\max}$ . Give the biochemical meaning for the intercepts with the horizontal and vertical axes.

**NB** The horizontal axis intercept has a negative value. The portion of the line in the third quadrant is not biochemically reasonable and is generally shown by a dotted line.