## Quick Tour of Mathcad and Examples

Mathcad provides a unique and powerful way to work with equations, numbers, tests and graphs.

## Features

Arithmetic
Functions
Plot functions
Define you own variables and functions
Visualize data in 2-D and 3-D
Compute sums and integrals
Do matrix computations
Solve equations
numerically
symbolically
Programming
Curve fitting and data analysis functions

Arithmetic, variables, constants, units, equal signs.
The $=$ sign means to print the value
The : sign means to define a variable [You type : but mathcad prints :=]
The $\sim$ sign means to define a variable
before the rest of the work sheet is evaluated

The <cntrl>=, $\quad=\mathbf{\imath}$, means to set equal "logically", that is without numerical evaluation.

The .. range operator defines the range in the toolbar or use a semicolon (; )

To separate expressions use the comma operator (, ).

Arithmetic, Functions, Calculus, Greek letters, Matrix, Symbolic, Programming etc ... can be selected in the Toolbar. (View, then Toolbar)
(a). Arithmetic To the right, type the following (NOT IN A TEXT REGION) [For tips on editing and typing expressions, see Editing below.] To add text Go to top menu and Insert and select Text Region

2/3=
2*3=
2^3=
compare
2/3:
(b). Now type the following:

$$
\begin{aligned}
& x: 2 \\
& y: 3 \\
& x / y= \\
& x^{\star} y= \\
& x^{\wedge} y=
\end{aligned}
$$

(c). Mathcad has several built-in constants. That is, Mathcad "knows" these constants.

Find the values of the following constants by typing

```
e=
\(p<\) Ctrl-g>=
```

[This means type p and hold down the Ctrl key while you press g.]

$$
\mathrm{R}=
$$

Note that R is NOT the gas constant.
Editing: Correcting a typing error: place the mouse just to the right of the wrong character or symbol.
Click and press the delete (back arrow) key. Try this: change 3 to 9 in the math expressions you created at left.

## Editing:

Whole mathcad expressions or text regions can be removed. Just select the desired region and press the delete or backspace key. There are two ways to select a region. Either click and drag into the region, or click once in the region and press <up arrow> repeating until whole region is outlined.

## Editing:

Expressions can be edited for variables and constants or for operators. To remove (and replace) and operator, click on the expression and use the up-arrow or down-arrow to select the part of the expression that contains the offending operator. (blue editing line) Press the back-arrow to remove it and press the desired operator key to replace it.

## Arithmetic:

$$
\frac{197}{13}=15.154
$$

$$
13^{7}=6.275 \times 10^{7}
$$

$$
\sqrt{\frac{1.837 \cdot 10^{2}}{50+6^{5}}}=0.153
$$

$$
\prod_{j=1}^{5} \sum_{i=1}^{j} \frac{1}{i}=13.082
$$

Functions:

$$
\log (145) \cdot \cos \left(\frac{3}{5} \cdot \pi\right)=-0.668
$$

## Complex numbers

$$
(5.1+4 i)^{2}+e^{3-2 i}=1.651+22.536 i
$$

Units: (to check or add new units use top menu Insert and Units)

$$
\frac{1250 \cdot \mathrm{~km}}{1.5 \cdot \mathrm{hr}}=231.481 \frac{\mathrm{~m}}{\mathrm{~s}} \quad \text { change units directly to } \mathrm{mph}
$$

## Function and Plot:

$$
z:=0,0.5 . .4 \quad f(z):=\sin (z)
$$

| $\mathrm{Z}=$ |
| :--- |
| 0.000 |
| 0.500 |
| 1.000 |
| 1.500 |
| 2.000 |
| 2.500 |
| 3.000 |
| 3.500 |
| 4.000 |

$\mathrm{f}(\mathrm{z})=$

| 0.000 |
| ---: | ---: |
| 0.479 |
| 0.841 |
| 0.997 |
| 0.909 |
| 0.598 |
| 0.141 |
| -0.351 |
| -0.757 |



## Variables and your own functions:

$$
\mathrm{a}:=4 \quad \mathrm{a} \cdot 5+\sqrt[3]{\mathrm{a}}=21.587
$$

Or $\quad f(x):=\frac{\sin (x)}{\frac{x}{a^{3}}}$

$$
f(1)=53.854
$$

$$
f(10)=-3.482
$$

## Calculus:

$$
\sum_{n=0}^{12} \frac{1}{n!}=2.718 \quad \int_{0}^{1} \frac{1}{1+\sqrt[3]{x}} d x=0.579
$$

## Matrix

$$
\left.\begin{array}{c}
A:=\left(\begin{array}{ccc}
4 & 5 & 1 \\
5 & 0 & -2 \\
-1 & 2 & 8
\end{array}\right) \\
A \cdot A^{-1}=\left(\begin{array}{cc}
1.000 & -6.939 \times 10^{-17} \\
1.388 \times 10^{-17} & -2.776 \times 10^{-17} \\
5.551 \times 10^{-17} & 0.000
\end{array} \quad 0.000\right.
\end{array}\right) \quad A^{-1}=\left(\begin{array}{ccc}
-0.024 & 0.232 & 0.061 \\
0.232 & -0.201 & -0.079 \\
-0.061 & 0.079 & 0.152
\end{array}\right)
$$

Solving Equations:(numerically, symbolically, using Symbolics, Root, Polyrrot and Given...Find build in functions)

$$
\begin{aligned}
& \text { Numericall. } \quad t:=3 \\
& \operatorname{root}\left(\mathrm{t}^{2}-\cosh (\mathrm{t}), \mathrm{t}\right)=2.594
\end{aligned}
$$

Giver

$$
\mathrm{v}:=1
$$

$$
v^{2}+3 v-5=0
$$

$\operatorname{Find}(v)=1.193$

Symbolically using "solve"

$$
x 1+1=\frac{1}{x 1} \text { solve, } x 1 \rightarrow\binom{\frac{1}{2} \cdot 5^{\frac{1}{2}}-\frac{1}{2}}{\frac{-1}{2}-\frac{1}{2} \cdot 5^{\frac{1}{2}}}
$$

$$
\sum_{i=1}^{n} i^{2} \rightarrow \frac{1}{3} \cdot(n+1)^{3}-\frac{1}{2} \cdot(n+1)^{2}+\frac{1}{6} \cdot n+\frac{1}{6}
$$

change the value of $\boldsymbol{n}$ in the Sum symbol

## Precautions

1. Numerical formatting: Under Format and Result there are three parameters that you have to be careful about.
(a) Under Number Format--General (Number of Decimal Places and Exponential Threshold)

Number of decimal places determines the number of figures displayed after the decimal (3 default).
Exponential Threshold determines how large or how small a number can become before the program switches to exponential notation (3 default)
(b) under Number Format , Tolerence determines the size of a number that will be displayed as 0
decimal $=9$ and
zero threshold = 9

$$
h 3:=h^{3} \mathrm{~h}:=6.626075 \cdot 10^{-34} \mathrm{~h} 3=2.909 \times 10^{-100}
$$

then change
zero threshold to 150

$$
10^{10} \mathrm{~h} 3=2.909 \times 10^{-90}
$$

Largest and smallest numbers are 10307 and 10-307

## 2. Errors:

$R$ is Rankine-scale degree temperature NOT the gas constant g : acceleration of gravity NOT gram Second is sec NOT s
Meter (unit of lenght SI system of unit) is m . In general you should not use ( m , sec, gm and K) as variable names.

Kelvin temperature is denoted K therefore if you need to use it as equilibrium constant add a subscript K eq for instance.

Mathcad evaluates expression from left to right and top to bottom.
Each sheet is indenpendent of the others.
Default numbering system for subscripts begins with 0 not 1. $x_{0}, x_{1}, \ldots$

## Fitting Functions

## I. Linear Regression:

Suppose the data consists of paired points $\left\{x_{i}\right\},\left\{y_{i}\right\}$. The regression refers to finding the coefficient $c_{j}$ of the following equation and to minimize the sum of squares of the deviations of the calculated and observed $y$.

$$
y=\sum_{i}\left[c_{i} \cdot f_{i}(x)\right]
$$

where $\mathrm{c}_{\mathrm{i}}$ and $\mathrm{f}_{\mathrm{i}}(\mathrm{x})$ are the coefficient and arbitrary functions of x , respectively.
"Fitting data to a linear function" of the form $y=a x+b$ where $a$ is the slope and $b$ the intercept.

## Example 1:

You need first to define the number of points N and then the range variable i (index):

$$
N:=10 \quad i:=0 . . N
$$

Data:

$$
x_{i}:=\quad y_{i}:=
$$

$$
\mathrm{a}:=\operatorname{slope}(\mathrm{x}, \mathrm{y}) \quad \mathrm{a}=1.927
$$

| -5 |
| :--- |
| -4 |
| -3 |
| -2 |
| -1 |
| 0 |
| 1 |
| 2 |
| 3 |
| 4 |
| 5 |$\quad$| -11 |
| :--- |
| -8 |
| -6.5 |
| -4 |
| -1.3 |
| 0 |
| 1.1 |
| 4 |
| 6.2 |
| 7 |
| 8.1 |

$$
\mathrm{b}:=\operatorname{intercept}(\mathrm{x}, \mathrm{y}) \quad \mathrm{b}=-0.400
$$

$$
\operatorname{fit}(z):=a \cdot z+b
$$

$$
z:=-6,-4.5 . .6
$$



## II. Least Squares Fits:

Suppose the dataset is $\left\{\left(x_{i}, y_{i}\right)\right.$ with $\left.i=1,2,3, \ldots . n\right\}$. And you need to find the best polynomial fit to these data points in the form of $f(z)$ :

$$
f(z)=\sum_{j=0}^{d}\left(c_{j} \cdot z^{j}\right) \quad \underset{\text { with } d=3}{\text { equivalent to }} \quad f(z)=c_{0}+c_{1} \cdot z+c_{2} \cdot z^{2}+c_{3} \cdot z^{3}
$$

The best fit of $y(x)$ to the data is found by minimizing the squared deviation function:
$M=\sum_{i=1}^{n}\left[y_{i}-\sum_{j=0}^{d}\left[c_{j} \cdot\left(x_{i}\right)^{j}\right]^{2} \quad \begin{array}{c}\text { equivalent to } \\ M=\left(y_{1}-c_{0}+c_{1} \cdot x_{1}\right)^{2}+\left(y_{2}-c_{0}+c_{1} \cdot x_{2}\right)^{2} \\ \text { with } n=2 \text { and } d=1\end{array}\right.$

Define a function $X$ such as $\quad X_{i}, j=\left(x_{i}\right)^{j}$

$$
\text { and a matrix } M \text { such as } \quad M=X^{\top} \cdot X
$$

Then the values of $c_{j}$ that minimize $M$ are given by

$$
\mathrm{c}=\mathrm{M}^{-1} \cdot \mathrm{X} \cdot \mathrm{y}
$$

Example 2: $\quad \mathrm{x}:=\left(\begin{array}{lllllll}-2.5 & -1.0 & -0.2 & 0.5 & 2.2 & 3.9 & 5.6\end{array}\right)^{\top}$

$$
y:=\left(\begin{array}{lllllll}
5.2 & 2.9 & 1.1 & 1.6 & 5.75 & 18 & 29
\end{array}\right)^{\top}
$$

$\mathrm{n}:=\operatorname{length}(\mathrm{x}) \quad$ or $\quad \mathrm{n}:=\operatorname{rows}(\mathrm{x}) \quad \mathrm{i}:=0 . . \mathrm{n}-1 \quad \mathrm{~d}:=2$

$$
\mathrm{j}:=0 . . \mathrm{d}
$$

Using the functions defined above $\quad X_{i, j}:=\left(x_{i}\right)^{j} \quad$ and $\quad M:=X^{\top} \cdot X$

$$
\begin{aligned}
& M=\left(\begin{array}{ccc}
7.000 & 8.500 & 58.950 \\
8.500 & 58.950 & 229.075 \\
58.950 & 229.075 & 1.278 \times 10^{3}
\end{array}\right) \\
& \mathrm{M}^{-1}=\left(\begin{array}{ccc}
0.247 & 0.029 & -0.017 \\
0.029 & 0.059 & -0.012 \\
-0.017 & -0.012 & 3.679 \times 10^{-3}
\end{array}\right)
\end{aligned}
$$

example: $3 \times 3$ matrix $A^{\top}=\left(\begin{array}{ccc}4.000 & 5.000 & 1.000 \\ 5.000 & 0.000 & -2.000 \\ -1.000 & 2.000 & 8.000\end{array}\right) \quad \mathrm{A}^{-1}=\left(\begin{array}{ccc}-0.024 & 0.232 & 0.061 \\ 0.232 & -0.201 & -0.079 \\ -0.061 & 0.079 & 0.152\end{array}\right)$
$\mathrm{A}^{\top}=\left(\begin{array}{ccc}4.000 & 5.000 & -1.000 \\ 5.000 & 0.000 & 2.000 \\ 1.000 & -2.000 & 8.000\end{array}\right) \quad|\mathrm{A}|=-164.000 \quad$ nonsingular (invertible) determinant is not zero

Back to our problem, the coefficients c are given by

$$
\begin{aligned}
& \underset{m}{\mathrm{c}:=\mathrm{M}^{-1} \cdot \mathrm{X}^{\top} \cdot \mathrm{y}} \quad \mathrm{c}=\left(\begin{array}{l}
1.681 \\
0.556 \\
0.798
\end{array}\right) \quad \underset{\mathrm{m}}{\mathrm{~g}} \mathrm{f}(\mathrm{z}):=\sum_{\mathrm{j}=0}^{\mathrm{d}}\left(\mathrm{c}_{\mathrm{j}} \cdot \mathrm{z}^{j}\right)
\end{aligned}
$$

Polynomial fitting function:

$$
Z:=-5,-4.9 . .8
$$



## Use of "linfit" function (MathCAD built in function)

Here $y(x)=c_{0}+c_{1} x+c_{2} x^{2}$. We need to define a function $Y(x)$ which gives the power of the variable $x$.

The linfit $(x, y, Y)$ function returns a vector containing the parameters used to create a linear combination of the functions in vector $Y$ that best approximates the data in $x$ and y in the least-squares sense.There must be at least as many data points as there are terms in Y .
$\mathrm{Y}(\mathrm{x})$ is a vector of functions; each element is one linear functional term in the fit function. In the case of a single linear function, Y is a scalar.

$$
\begin{aligned}
Y(x) & :=\left(\begin{array}{c}
1 \\
x \\
x^{2}
\end{array}\right) \quad c^{\prime}:=\operatorname{linfit}(x, y, Y) \quad c^{\prime}=\left(\begin{array}{c}
1.681 \\
0.556 \\
0.798
\end{array}\right) \quad z:=-5,-4.9 . .10 \\
\operatorname{fit}^{\prime}(z) & :=Y(z) \cdot \operatorname{linfit}(x, y, Y)
\end{aligned}
$$



More functions are defined in MathCAD

1. General least square such as genfit, linfit , regress with interp.
2. Specialized least square functions such expfit, Infit, logfit, pwrfit, sinfit etc ...

## Logarithmic Regression

The use of Mathcad's linfit function to fit data to a logarithmic model is illustated bellow. Two specialized built-in fitting functions, logfit and Infit, may also be used.
The following Data Table contains a set of data that may be best approximated by a logarithmic model function.

$$
\begin{aligned}
& \begin{array}{l}
\text { data }:=\begin{array}{|l|r|r|}
\hline & 0 & 1 \\
\hline 0 & 1 & 4.18 \\
\hline 1 & 2 & 4.67 \\
\hline 2 & 3 & 5.3 \\
\hline 3 & 4 & 5.37 \\
\hline 4 & 5 & 5.45 \\
\hline 5 & 6 & 5.74 \\
\hline 6 & 7 & 5.65 \\
\hline 7 & 8 & 5.84 \\
\hline 7 & 9 & 6.36 \\
\hline 8 & 10 & 6.38 \\
\hline 9 &
\end{array}
\end{array} \\
& X:=\operatorname{data}^{\langle 0\rangle} \quad Y:=\operatorname{data}^{\langle 1\rangle}
\end{aligned}
$$

Enter the vector of functions to fit. In this example, we fit this data to the model function $\mathrm{y}=\mathrm{a} \cdot \ln (\mathrm{x})+\mathrm{b} \cdot \sqrt{\mathrm{x}}+\mathrm{c}$, where $\mathrm{a}, \mathrm{b}$, and c are unknown.

$$
F(x):=\left(\begin{array}{c}
\ln (x) \\
\sqrt{x} \\
1
\end{array}\right)
$$

1. Call the linfit function using the data and the vector of functions, $F$.

$$
\text { S1 := linfit }(X, Y, F)
$$

Define a function that uses these newly found parameter values in the logarithmic model. Also, define a range variable over which to graph the function.

$$
\mathrm{f}_{\mathrm{m}}(\mathrm{x}):=\mathrm{F}(\mathrm{x}) \cdot \mathrm{S} 1 \quad \mathrm{z}:=\min (\mathrm{X}) . . \max (\mathrm{X}) \quad \text { (range variable for plotting) }
$$

A graph of the model function with the newly found parameter values and the original data points reveals a good fit.

$\operatorname{corr}(\overrightarrow{f(X)}, Y)=0.979 \quad$ correlation coefficient
2. A simpler way to perform a customized log fit is by using the Infit function. The data is fitted with the model function $y=a \cdot \ln (x)+b:$

$$
\begin{aligned}
& \mathrm{T} 1:=\operatorname{lnfit}(\mathrm{X}, \mathrm{Y}) \quad \mathrm{T} 1=\binom{0.881}{4.173} \\
& \mathrm{~h}(\mathrm{x}):=\mathrm{T} 1_{0} \cdot \ln (\mathrm{x})+\mathrm{T} 1_{1}
\end{aligned}
$$

$$
\mathrm{Z}:=0 . . \operatorname{last}(\mathrm{X}) \quad \text { (range variable for plotting), a different way to set a range of values }
$$



$$
\operatorname{corr}(\overrightarrow{\mathrm{h}(\mathrm{X})}, \mathrm{Y})=0.979 \quad \text { correlation coefficient }
$$

3. You can also use Mathcad's logfit function, which fits data to the model function. The model function used here is $y=a \cdot \ln (x+b)+c$.

$$
\begin{aligned}
& \mathrm{Q}:=\left(\begin{array}{l}
1 \\
0 \\
4
\end{array}\right) \quad \text { guess vector } \\
& \mathrm{M}:=\operatorname{logfit}(\mathrm{X}, \mathrm{Y}, \mathrm{Q}) \quad \mathrm{M}=\left(\begin{array}{c}
0.864 \\
-0.097 \\
4.227
\end{array}\right) \\
& \mathrm{k}(\mathrm{x}):=\mathrm{M}_{0} \cdot \ln \left(\mathrm{x}+\mathrm{M}_{1}\right)+\mathrm{M}_{2}
\end{aligned}
$$

$$
z:=0 . . \operatorname{last}(X) \quad \text { range variable for plotting }
$$



## III. Polynomial Regression:

## Polynomial function that best fit the set of data points

The function regress $(\mathbf{X}, \mathbf{Y}, \boldsymbol{k})$ returns a vector $\boldsymbol{s}$ which interp uses to find the $\boldsymbol{k}^{\text {th }}$ order polynomial that best fits the $\boldsymbol{x}$ and $\boldsymbol{y}$ data values.

The function interp(s,X,Y,X) returns interpolated $\mathbf{y}$ value corresponding to $\boldsymbol{x}$.
$X$ is a vector of real data values in ascending order (the $\boldsymbol{x}$ values).
$Y$ is a vector of real data values (the $\boldsymbol{y}$ values).
$s$ is a vector generated by the function regress.
$\boldsymbol{k}$ is a positive integer specifying the order of the polynomial you want to use to fit the data. Usually you'll want $\boldsymbol{k}<5$.
$\mathbf{x}$ is the value of the independent variable at which you want to evaluate the regression curve.

## Notes:

-Regress is useful when you have a set of measured $\boldsymbol{y}$ values corresponding to $\boldsymbol{x}$ values and you want to fit a single polynomial of any order to those $\boldsymbol{y}$ values.

- You should always use interp after using the regress function.
- Since regress tries to accommodate all your data points using a single polynomial, it will not work well when your data does not behave like a single polynomial.

Example : Consider a matrix of $X-Y$ data to be analyzed ( $X$ coordinate in first column, $Y$ coordinate in second)

$$
\text { data }:=\left(\begin{array}{cc}
0 & 9.1 \\
1 & 7.3 \\
2 & 3.2 \\
3 & 4.6 \\
4 & 4.8 \\
5 & 2.9 \\
6 & 5.7 \\
7 & 7.1 \\
8 & 8.8 \\
9 & 10.2
\end{array}\right) \quad \begin{aligned}
& \mathrm{X}:=\operatorname{data}^{\langle 0\rangle} \\
& \mathrm{Y}:=\operatorname{data}^{\langle 1\rangle} \\
& \\
& \\
& \\
& \\
&
\end{aligned}
$$

Enter the degree of polynomial to fit, value of $k$ : $\quad \underset{\sim}{k}:=3$
Define the polynomial function that fit best the set of data points using "regress" as follows:

$$
\mathrm{s}_{1}:=\operatorname{regress}(\mathrm{X}, \mathrm{Y}, \mathrm{k}) \quad \mathrm{s}_{1}=\left(\begin{array}{c}
3.000 \\
3.000 \\
3.000 \\
9.298 \\
-3.438 \\
0.609 \\
-0.024
\end{array}\right)
$$

Use the function "interp" as the polynomial fitting function:
$\begin{array}{ll}\text { fit }_{1}(x):=\operatorname{interp}\left(s_{1}, X, Y, x\right) & \begin{array}{l}\text { interpolation of } y \text { value corresponding to } x \\ \text { vector } s\end{array}\end{array}$

$$
x:=0,1 . .10
$$


coeffs $:=$ submatrix $\left(s_{1}, 3\right.$, length $\left.\left(s_{1}\right)-1,0,0\right)$
The coefficients are given by: $\quad$ coeffs $=\left(\begin{array}{c}9.298 \\ -3.438 \\ 0.609 \\ -0.024\end{array}\right)$

$$
\text { coeffs }^{\top}=\left(\begin{array}{llll}
9.298 & -3.438 & 0.609 & -0.024
\end{array}\right) \quad \text { (matrix transpose) }
$$

$$
\operatorname{Fit}(x):=\sum_{i=0}^{k}\left(\text { coeffs }_{i} \cdot x_{i}\right)
$$

cspline, pspline and lspline and interp functions for connecting $\mathrm{X}-\mathrm{Y}$ data.
Enter a matrix of X-Y data to be interpolated:

M $_{1}:=$|  | 0 | 1 |
| ---: | ---: | ---: |
| 0 | 0 | 0 |
| 1 | 1 | 3 |
| 2 | 4 | 27.57 |
| 3 | 5 | 25 |
| 4 | 7 | 18 |
| 5 | 8 | 30.41 |
| 6 | 11 | 48 |
| 7 | 13 | 62 |
| 8 | 14 | 59.89 |
| 9 | 16 | 72.28 |
| 10 | 17 | 83 |
| 11 | 18 | 65 |
| 12 | 19 | 78 |
| 13 | 20 | 85 |
| 14 |  |  |

$$
M_{1}:=\operatorname{csort}\left(M_{1}, 0\right) \quad X:=M_{1}{ }^{\langle 0\rangle} \quad Y:=M_{1}{ }^{\langle 1\rangle}
$$

Spline coefficients:

$$
S_{1}:=\text { Ispline }(X, Y)
$$

Sample interpolated values:

$$
\text { fit }(3)=21.641
$$

$$
\text { fit }(9.5)=41.231
$$

Fitting function:

$$
\operatorname{fit}^{\mathrm{fit}}(\mathrm{x}):=\operatorname{interp}\left(\mathrm{S}_{1}, \mathrm{X}, \mathrm{Y}, \mathrm{x}\right)
$$

$$
z:=0,0.25 . .20
$$



Any ot the fiiting functions can then be used for calculations (integrals, derivatives etc ...)

## Multivariate Regression using regress

Assume you have the following data. In this example, $X$ and $Y$ are the independent variables, and $Z$ is the dependent variable.

$$
\mathrm{i}:=0 . .14
$$

$$
X_{i}:=\quad Y_{i}:=\quad Z_{i}:=
$$

| 47 | 20 | 49 |
| :---: | :---: | :---: |
| 80 | 28 | 118 |
| 125 | 48 | 145 |
| 31 | 8 | 20 |
| 116 | 35 | 81 |
| 99 | 40 | 68 |
| 43 | 13 | 51 |
| 30 | 2 | 13 |
| 150 | 24 | 80 |
| 112 | 20 | 43 |
| 42 | 6 | 27 |
| 150 | 47 | 71 |
| 92 | 29 | 101 |
| 89 | 23 | 46 |
| 46 | 15 | 27 |

The higher dimensional form of regress takes three arguments:
a real array, $\mathrm{M}_{\mathbf{2}}$, each column of which represents data corresponding to one of the independent variables
a real vector, V , representing the data for the dependent variable a positive integer, n , specifying the degree of the polynomial function to which the data will be fit

$$
\begin{array}{ll}
\mathrm{M}_{2}:=\operatorname{augment}(\mathrm{X}, \mathrm{Y}) & \begin{array}{l}
\text { Combine the independent variables into a single matrix using the } \\
\text { augment function }
\end{array}
\end{array}
$$

$$
\mathrm{n}:=2
$$

Define the degree of the polynomial. (This value could also be passed directly to regress.)

Define the regress function, and view the results.

$$
\mathrm{S}_{2}:=\operatorname{regress}\left(\mathrm{M}_{2}, \mathrm{Z}, \mathrm{n}\right)
$$

$\mathrm{S}_{2}=\left(\begin{array}{c}3.000 \\ 3.000 \\ 2.000 \\ -0.056 \\ 0.092 \\ 2.530 \\ -14.760 \\ 0.691 \\ 2.980 \times 10^{-3}\end{array}\right)$

## Defining a function using interp

You can use the interp function to define a two-variable polynomial function which uses these coefficients. interp is able to interpret the order of the coefficients and the monomial terms with which they are associated. In particular, it uses the first three terms of the vector, $\mathrm{S}_{2}$, to determine the built-in function being called (3 for regress), the vector position of the first coefficient (always 3 ), and the degree of the fitting function ( $n$ ).
$f_{m}(x, y):=\operatorname{interp}\left[S_{2}, M_{2}, Z,\binom{x}{y}\right]$
This definition will vary depending upon the number of independent variables. In this example, we have two independent variables, $x$ and $y$.

Call the function for specific values but do not use the function to extrapolate. It should be used for interpolation only.

$$
f(90,12)=54.958 \quad \text { This function call will vary depending upon the number of }
$$ independent variables you have.

Here we are dealing with a 2D case of multivariate regression. Therefore, we can graph the result as a surface over ranges for $x$ and $y$.

$$
F:=\text { CreateMesh }(f, \min (X), \max (X), \min (Y), \max (Y), 20,20)
$$



The regression surface passes through the space minimizing the distance to each of the data points, shown in black on the graph.

$$
F,(X, Y, Z)
$$

## Quantum Mechanics

Construction of the double well potential $\mathrm{V} 2(\mathrm{x})$ and calculation of the ten first energy levels of the system. Plot of the wavefunctions and probability densities. Tunneling effect.

$$
V 2(x 1):=\left\lvert\, \begin{array}{lll}
0 & \text { if } & 0<x 1<2 \\
20 & \text { if } & 2 \leq x 1<4 \\
0 & \text { if } & 4 \leq x 1<6 \\
60 & \text { otherwise }
\end{array} \quad x 1\right.:=-1,-0.999 . .10
$$

Differential Equation to solve

$$
\begin{aligned}
& \mathrm{eV}:=\mathrm{e} \cdot \mathrm{voll} \quad \mathrm{eV} \equiv 1.602177 \cdot 10^{-19} \cdot \mathrm{~J} \\
& \mathrm{n}_{1}:=0 . .9 \quad \mathrm{~h}:=6.626 \cdot 10^{-34} \cdot \mathrm{~J} \cdot \mathrm{~s} \quad \mathrm{a}:=5 \cdot 10^{-9} \cdot \mathrm{~m} \quad \mathrm{~m}:=9.1 \cdot 10^{-31} \cdot \mathrm{~kg} \\
& \mathrm{E} 1\left(\mathrm{n}_{1}\right):=\frac{\mathrm{h}^{2} \cdot \mathrm{n}_{1}^{2}}{8 \cdot \mathrm{~m} \cdot \mathrm{a}^{2}} \mathrm{E} 1\left(\mathrm{n}_{1}\right)= \\
& \begin{array}{|c|}
\hline 0.000 \\
\hline 0.015 \\
\hline 0.060 \\
\hline 0.136 \\
\hline 0.241 \\
\hline 0.376 \\
\hline 0.542 \\
\hline 0.738 \\
\hline 0.964 \\
\hline 1.220 \\
\hline
\end{array}
\end{aligned}
$$

$\varepsilon 1:=15 \quad$ change the values of $\varepsilon$ until to get tunneling about 18

$$
y 1:=\binom{0}{1} \quad D 1(x 1, y 1):=\left[\begin{array}{c}
y 1_{1} \\
(\mathrm{~V} 2(\mathrm{x} 1)-\varepsilon 1) \cdot \mathrm{y} 1_{0}
\end{array}\right]
$$

$$
\underset{w}{v}:=\operatorname{rkfixed}(\mathrm{y} 1,0,6,400, \mathrm{D} 1)
$$




No tunneling is observed for $\varepsilon$ less than 15. the particle tunnels through the potential barrier at $\varepsilon$ greater than 16. At $\varepsilon$ greater than 16, tunneling takes place and a wavefunction appears after $x=-2$.

