## Taylor Expansions in 2d

In your first year Calculus course you developed a family of formulae for approximating a function $F(t)$ for $t$ near any fixed point $t_{0}$. The crudest approximation was

$$
F\left(t_{0}+\Delta t\right) \approx F\left(t_{0}\right)
$$

The next better approximation included a correction that is linear in $\Delta t$.

$$
F\left(t_{0}+\Delta t\right) \approx F\left(t_{0}\right)+F^{\prime}\left(t_{0}\right) \Delta t
$$

The next better approximation included a correction that is quadratic in $\Delta t$.

$$
F\left(t_{0}+\Delta t\right) \approx F\left(t_{0}\right)+F^{\prime}\left(t_{0}\right) \Delta t+\frac{1}{2} F^{\prime \prime}\left(t_{0}\right) \Delta t^{2}
$$

And so on. The approximation that includes all corrections up to order $\Delta t^{n}$ is

$$
F\left(t_{0}+\Delta t\right) \approx F\left(t_{0}\right)+F^{\prime}\left(t_{0}\right) \Delta t+\frac{1}{2!} F^{\prime \prime}\left(t_{0}\right) \Delta t^{2}+\frac{1}{3!} F^{(3)}\left(t_{0}\right) \Delta t^{3}+\cdots+\frac{1}{n!} F^{(n)}\left(t_{0}\right) \Delta t^{n}
$$

You may have also found a formula for the error introduced in making this approximation. The error $E_{n}(\Delta t)$ is defined by

$$
F\left(t_{0}+\Delta t\right)=F\left(t_{0}\right)+F^{\prime}\left(t_{0}\right) \Delta t+\frac{1}{2!} F^{\prime \prime}\left(t_{0}\right) \Delta t^{2}+\cdots+\frac{1}{n!} F^{(n)}\left(t_{0}\right) \Delta t^{n}+E_{n}(\Delta t)
$$

and obeys

$$
E_{n}(\Delta t)=\frac{1}{(n+1)!} F^{(n+1)}\left(t^{*}\right) \Delta t^{n+1}
$$

for some (unknown) $t^{*}$ between $t_{0}$ and $t_{0}+\Delta t$. Even though we do not know what $t^{*}$ is, we can still learn a lot from this formula for $E_{n}$. If we know that $\left|F^{(n+1)}(t)\right| \leq M_{n+1}$ for all $t$ between $t_{0}$ and $t_{0}+\Delta t$, then $\left|E_{n}(\Delta t)\right| \leq \frac{M_{n+1}}{(n+1)!} \Delta t^{n+1}$, which tells us that $E_{n}(\Delta t)$ goes to zero like the $(n+1)^{\text {st }}$ power of $\Delta t$ as $\Delta t$ tends to zero.

We now generalize all this to functions of more than one variable. To be concrete and to save writing, we'll just look at functions of two variables, but the same strategy works for any number of variables. Suppose that we wish to approximate $f\left(x_{0}+\Delta x, y_{0}+\Delta y\right)$ for $\Delta x$ and $\Delta y$ near zero. The trick is to write

$$
f\left(x_{0}+\Delta x, y_{0}+\Delta y\right)=F(1) \text { with } F(t)=f\left(x_{0}+t \Delta x, y_{0}+t \Delta y\right)
$$

and think of $x_{0}, y_{0}, \Delta x$ and $\Delta y$ as constants so that $F$ is a function of the single variable $t$. Then we can apply our single variable formulae with $t_{0}=0$ and $\Delta t=1$. To do so we need to compute various derivatives of $F(t)$ at $t=0$, by applying the chain rule to

$$
F(t)=f(x(t), y(t)) \text { with } x(t)=x_{0}+t \Delta x, y(t)=y_{0}+t \Delta y
$$

Since $\frac{d}{d t} x(t)=\Delta x$ and $\frac{d}{d t} y(t)=\Delta y$, the chain rule gives

$$
\begin{aligned}
\frac{d F}{d t}(t)= & \frac{\partial f}{\partial x}(x(t), y(t)) \frac{d}{d t} x(t)+\frac{\partial f}{\partial y}(x(t), y(t)) \frac{d}{d t} y(t) \\
= & \frac{\partial f}{\partial x}\left(x_{0}+t \Delta x, y_{0}+t \Delta y\right) \Delta x+\frac{\partial f}{\partial y}\left(x_{0}+t \Delta x, y_{0}+t \Delta y\right) \Delta y \\
\frac{d^{2} F}{d t^{2}}(t)= & {\left[\frac{\partial f_{x}}{\partial x}(x(t), y(t)) \frac{d}{d t} x(t)+\frac{\partial f_{x}}{\partial y}(x(t), y(t)) \frac{d}{d t} y(t)\right] \Delta x } \\
& +\left[\frac{\partial f_{y}}{\partial x}(x(t), y(t)) \frac{d}{d t} x(t)+\frac{\partial f_{y}}{\partial y}(x(t), y(t)) \frac{d}{d t} y(t)\right] \Delta y \\
= & \frac{\partial^{2} f}{\partial x^{2}} \Delta x^{2}+\frac{\partial^{2} f}{\partial y \partial x} \Delta y \Delta x+\frac{\partial^{2} f}{\partial x \partial y} \Delta x \Delta y+\frac{\partial^{2} f}{\partial y^{2}} \Delta y^{2} \\
= & \frac{\partial^{2} f}{\partial x^{2}} \Delta x^{2}+2 \frac{\partial^{2} f}{\partial x \partial y} \Delta x \Delta y+\frac{\partial^{2} f}{\partial y^{2}} \Delta y^{2}
\end{aligned}
$$

and so on. It's not hard to prove by induction that, in general,

$$
F^{(n)}(t)=\sum_{m=0}^{n}\binom{n}{m} \frac{\partial^{n}}{\partial x^{m} \partial y^{n-m}}\left(x_{0}+t \Delta x, y_{0}+t \Delta y\right) \Delta x^{m} \Delta y^{n-m}
$$

where $\binom{n}{m}=\frac{n!}{m!(n-m)!}$ is the standard binomial coefficient. So when $t=0$,

$$
\begin{aligned}
F(0) & =f\left(x_{0}, y_{0}\right) \\
\frac{d F}{d t}(0) & =\frac{\partial f}{\partial x}\left(x_{0}, y_{0}\right) \Delta x+\frac{\partial f}{\partial y}\left(x_{0}, y_{0}\right) \Delta y \\
\frac{d^{2} F}{d t^{2}}(0) & =\frac{\partial^{2} f}{\partial x^{2}}\left(x_{0}, y_{0}\right) \Delta x^{2}+2 \frac{\partial^{2} f}{\partial x \partial y}\left(x_{0}, y_{0}\right) \Delta x \Delta y+\frac{\partial^{2} f}{\partial y^{2}}\left(x_{0}, y_{0}\right) \Delta y^{2}
\end{aligned}
$$

Subbing these into

$$
f\left(x_{0}+\Delta x, y_{0}+\Delta y\right)=\left.F\left(t_{0}+\Delta t\right)\right|_{t_{0}=0, \Delta t=1}=F(0)+F^{\prime}(0)+\frac{1}{2} F^{\prime \prime}(0)+\cdots
$$

gives

$$
\begin{aligned}
f\left(x_{0}+\Delta x, y_{0}+\Delta y\right)= & f\left(x_{0}, y_{0}\right) \\
& +f_{x}\left(x_{0}, y_{0}\right) \Delta x+f_{y}\left(x_{0}, y_{0}\right) \Delta y \\
& +\frac{1}{2}\left[f_{x x}\left(x_{0}, y_{0}\right) \Delta x^{2}+2 f_{x y}\left(x_{0}, y_{0}\right) \Delta x \Delta y+f_{y y}\left(x_{0}, y_{0}\right) \Delta y^{2}\right] \\
& +O\left(\Delta x^{3}+\Delta y^{3}\right)
\end{aligned}
$$

