## Transform Exp Data for LinReg

## September 18, 2017

```
In [4]: import numpy as np
        import matplotlib
        import matplotlib.pyplot as plt
        %matplotlib inline
        # real params
        z_0 = 10
        alpha = 3
        # data vectors
        n = 500
        t = np.linspace(0, 2, n)
        z = z_0*np.exp(-alpha*t + np.random.normal(0, z_0/20, n))
        def fit_linear(x,y):
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            Given vectors of data points (x,y), performs a fit for the linear model:
               yhat = beta0 + beta1*x,
            The function returns beta0, beta1 and rsq, where rsq is the coefficient of determination.
            xmean = np.mean(x)
            ymean = np.mean(y)
            Sxx = np.sum((x-xmean) ** 2)
            Sxy = np.sum((x-xmean) * (y-ymean))
            Syy = np.sum((y-ymean) ** 2)
            beta1 = Sxy / Sxx
            beta0 = ymean - beta1*xmean
            SStot = Syy
            def predict(x):
                return beta0 + beta1*x
            SSres = np.sum((y - predict(x)) ** 2)
            rsq = 1 - SSres/SStot
            return beta0, beta1, rsq
        # setup plotting area
```

```
fig, (tplt, xplt, x_tplt) = plt.subplots(1, 3, figsize=(16,6), sharex='col', sharey='row')
# plot original data
tplt.plot(t, z, 'o')
tplt.grid(True)
tplt.set_xlabel('t')
tplt.set_ylabel('z(t) = z_0 * exp(-a * t)')
# fit original data
t_beta0, t_beta1, t_Rsq = fit_linear(t, z)
z_hat = t_beta0 + t_beta1*t
t_RSSn = np.sum((z - z_hat) ** 2) / len(z)
# plot fit
tplt.plot(t, z_hat, '-', linewidth=3)
tplt.text(1.25, 7, "$R^2 = $"+str(np.round(t_Rsq,2)), fontsize=15)
tplt.text(1.25, 4, "RSS = "+str(np.round(t_RSSn,2)), fontsize=15)
# transform data
x = t
y = np.log(z)
# plot transformed data
xplt.plot(x, y, 'o')
xplt.grid(True)
xplt.set_xlabel('x')
xplt.set\_ylabel('y(x) = log [ z_0 * exp( -a * t ) ]')
# fit transformed data
x_beta0, x_beta1, x_Rsq = fit_linear(x, y)
y_hat = x_beta0 + x_beta1*x
x_RSSn = np.sum((y - y_hat) ** 2) / len(y)
# plot fit
xplt.plot(x, y_hat,'-',linewidth=3)
xplt.text(1.25, 7, "$R^2 = $"+str(np.round(x_Rsq,2)), fontsize=15)
xplt.text(1.25, 4, "RSS = "+str(np.round(x_RSSn, 2)), fontsize=15)
# tranform back
t_t = x
t_z = np.exp(y)
t_z_{hat} = np.exp(y_{hat})
```

```
# plot in t, z space
        x_tplt.plot(t_t, t_z, 'o')
        x_tplt.plot(t_t, t_z_hat,'-',linewidth=3)
        x_{tplt.text}(1.25, 7, "$R^2 = $"+str(np.round(x_Rsq,2)), fontsize=15)
        x_{tplt.text(1.25, 4, "RSS = "+str(np.round(x_RSSn, 2)), fontsize=15)
        x_tplt.grid(True)
        x_tplt.set_xlabel('t')
        # check paramiter estimates against original distribution
        z_{hat} = np.exp(x_{beta0})
        alpha_hat = -x_beta1
        print("z_0 = {}\talpha
                                     = {}".format(z_0, alpha))
        print("z_hat = {:.2f}\talpha_hat = {:.2f}".format(z_hat, alpha_hat))
        print("% err = {:.2f}%\t % err = {:.2f}%"
                  .format(100*abs(z_hat-z_0)/z_0, 100*abs(alpha_hat-alpha)/alpha))
z_0 = 10
                     alpha
z_{hat} = 10.01
                     alpha_hat = 3.00
% err = 0.14%
                        % err = 0.16%
```

