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## Anexo 1. A Integral de Temperatura.

$$\xi = 1 - \exp(-\theta)^n \quad \text{onde } \theta = K(T) * t \quad (\text{A.1.1})$$

Assumindo que na expressão anterior  $T = \text{cte}$  durante um intervalo infinitesimal de tempo e diferenciando com relação a  $t$  :

$$d\theta = K(T) dt = K_o \exp\left(-\frac{E}{R*T}\right) dt \quad (\text{A.1.2})$$

Usando a relação  $\beta = \frac{dT}{dt}$ :

$$\int_{\theta(T_o)}^{\theta(T)} d\theta' = \theta(T) - \theta(T_o) = \int_{T_o}^T \frac{K_o}{\beta} \exp\left(-\frac{E}{R*T}\right) dT \quad (\text{A.1.3})$$

Propõe-se uma mudança de variável na integral da temperatura:  $T = \varphi(y) = \frac{E}{Ry}$  Cambio de variável em a integral definida:

$$\int_a^b f(x)dx = \int_{\alpha}^{\beta} f[\varphi(y)] \varphi'(y)dy \quad \text{onde } x = \varphi(y) \therefore y = \varphi^{-1}(x) \Rightarrow \alpha = \varphi^{-1}(a), \beta = \varphi^{-1}(b) \Rightarrow y = \frac{E}{RT} \therefore \alpha = \frac{E}{RT}|_{T_o}, \beta = \frac{E}{RT}|_T \quad (\text{A.1.4})$$

De esta forma, usando as relações (A.1.4) na equação (A.1.3) resultam:

$$\theta(T) - \theta(T_o) = \frac{K_o}{\beta} \int_{\alpha}^{\beta} \exp(-y) \left[\frac{E}{Ry}\right]' dy = -\frac{K_o E}{\beta R} \int_{\alpha}^{\beta} \frac{\exp(-y)}{y^2} dy \quad (\text{A.1.5})$$

Resolvendo a ultima integral por partes, teremos :

$$v = \exp(-y); dv = -\exp(-y) dy; du = \frac{1}{y^2} dy; u = -\frac{1}{y}$$

$$\Rightarrow \int_{\alpha}^{\beta} \frac{\exp(-y)}{y^2} dy = \left(-\frac{\exp(-y)}{y}\right)\Big|_{\alpha}^{\beta} - \int_{\alpha}^{\beta} \frac{\exp(-y)}{y} dy$$

Substituindo em (A.1.5) se tem:

$$\begin{aligned} \theta(T) - \theta(T_o) &= -\frac{K_o E}{\beta R} \left\{ -\frac{RT}{E} \exp\left(-\frac{E}{RT}\right) + \frac{RT_o}{E} \exp\left(-\frac{E}{RT_o}\right) - \int_{\alpha}^{\beta} \frac{\exp(-y)}{y} dy \right\} = \\ &= -\frac{K_o}{\beta} \left\{ -T \exp\left(-\frac{E}{RT}\right) + T_o \exp\left(-\frac{E}{RT_o}\right) - \frac{E}{R} \int_{\alpha}^{\beta} \frac{\exp(-y)}{y} dy \right\} \end{aligned} \quad (\text{A.1.6})$$

Lembrando as conhecidas relações [100,101]:

$$E_1(x) = \int_x^{\infty} \frac{\exp(-y)}{y} dy \quad E_2(x) = \int_1^{\infty} \frac{\exp(-xy)}{y^2} dy \quad E_2(x) = \exp(-x) - xE_1(x)$$

$$\exp(-x) = E_{2(x)} + xE_{1(x)} = \int_1^\infty \frac{\exp(-xy)}{y^2} dy + x \int_x^\infty \frac{\exp(-y)}{y} dy \quad (\text{A.1.7})$$

Considerando a  $x = E/RT$ ,  $x_0 = E/RT_0$  e substituindo (A.1.7) em (A.1.6):

$$\theta(T) - \theta(T_0) = -\frac{K_0}{\beta} \left[ -T \left\{ \int_1^\infty \frac{\exp(-xy)}{y^2} dy + x \int_x^\infty \frac{\exp(-y)}{y} dy \right\} + T_0 \left\{ \int_1^\infty \frac{\exp(-x_0 y)}{y^2} dy + x_0 \int_{x_0}^\infty \frac{\exp(-y)}{y} dy \right\} \right] - \frac{E}{R} \int_\alpha^\beta \frac{\exp(-y)}{y} dy \quad (\text{A.1.8})$$

Aplicando a propriedade distributiva e agrupando :

$$\theta(T) - \theta(T_0) = \frac{K_0}{\beta} \left\{ T \int_1^\infty \frac{\exp(-\frac{E}{RT} y)}{y^2} dy - T_0 \int_1^\infty \frac{\exp(-\frac{E}{RT_0} y)}{y^2} dy - \frac{E}{R} \int_\infty^{\frac{E}{RT}} \exp(-y) dy - \frac{E}{R} \int_{\frac{E}{RT_0}}^\infty \exp(-y) dy + \frac{E}{R} \int_\alpha^\beta \frac{\exp(-y)}{y} dy \right\}$$

Simplificando as ultimas três integrais, teremos :

$$\theta(T) - \theta(T_0) = \frac{K_0}{\beta} \left\{ T \int_1^\infty \frac{\exp(-\frac{E}{RT} y)}{y^2} dy - T_0 \int_1^\infty \frac{\exp(-\frac{E}{RT_0} y)}{y^2} dy \right\} = \frac{K_0}{\beta} \left[ T E_2\left(\frac{E}{RT}\right) - T_0 E_2\left(\frac{E}{RT_0}\right) \right] \quad (\text{A.1.9})$$

Substituindo a equação (A.1.7) em a equação (A.1.9)

$$\theta(T) - \theta(T_0) = \frac{K_0}{\beta} \left[ T \left( \exp\left(-\frac{E}{RT}\right) - \frac{E}{RT} E_1\left(\frac{E}{RT}\right) \right) - T_0 \left( \exp\left(-\frac{E}{RT_0}\right) - \frac{E}{RT_0} E_1\left(\frac{E}{RT_0}\right) \right) \right] \quad (\text{A.1.10})$$

Realizando a aproximação:

$$E_1(x) = \int_x^\infty \frac{\exp(-y)}{y} dy \cong \frac{\exp(-x)}{x} \left( 1 - \frac{1!}{x} + \frac{2!}{x^2} - \frac{3!}{x^3} + \dots \right)$$

$$\theta(T) - \theta(T_0) =$$

$$\frac{K_0}{\beta} \left\{ T \left[ \exp\left(-\frac{E}{RT}\right) - \frac{E}{RT} \left( \frac{\exp(-\frac{E}{RT})}{\frac{E}{RT}} \left( 1 - \frac{1!}{\frac{E}{RT}} + \frac{2!}{(\frac{E}{RT})^2} - \dots \right) \right) \right] - T_0 \left[ \exp\left(-\frac{E}{RT_0}\right) - \frac{E}{RT_0} \left( \frac{\exp(-\frac{E}{RT_0})}{\frac{E}{RT_0}} \left( 1 - \frac{1!}{\frac{E}{RT_0}} + \frac{2!}{(\frac{E}{RT_0})^2} - \dots \right) \right) \right] \right\}$$

Simplificando a ultima equação e considerando somente ate a segunda potencia no desenvolvimento em serie,teremos :

$$\theta(T) - \theta(T_o) = K_o / \beta \left\{ T \exp \left( -\frac{E}{RT} \right) \left( 1 - 1 + \frac{RT}{E} - 2 \left( \frac{RT}{E} \right)^2 \right) - T_o \exp \left( -\frac{E}{RT_o} \right) \left( 1 - 1 + \frac{RT_o}{E} - 2 \left( \frac{RT_o}{E} \right)^2 \right) \right\}$$

$$\theta(T) - \theta(T_o) = \frac{K_o T}{\beta} \exp \left( -\frac{E}{RT} \right) \left( \frac{RT}{E} - 2 \left( \frac{RT}{E} \right)^2 \right) - \frac{K_o T_o}{\beta} \exp \left( -\frac{E}{RT_o} \right) \left( \frac{RT_o}{E} - 2 \left( \frac{RT_o}{E} \right)^2 \right)$$

Assumindo que  $T_o \approx 0$  (na realidade  $T_o \ll T$ ), e para o qual  $\theta(T_o) \approx 0$ , (na realidade  $\theta(T_o) \ll \theta(T)$ ):

$$\theta(T) = \frac{T^2 R}{\beta E} \left( K_o \exp \left( -\frac{E}{RT} \right) \right) \left( 1 - 2 \frac{RT}{E} \right) \quad (\text{A.1.11})$$

## Anexo 2.

Tomando a equação (3.7)

$$\frac{d\xi}{dT} = \frac{1}{P_1 - P_o} \left\{ \left( \frac{dP}{dT} - \frac{dP_o}{dT} \right) - \frac{P - P_o}{P_1 - P_o} \left( \frac{dP_1}{dT} - \frac{dP_o}{dT} \right) \right\} \quad (A.2.1)$$

Derivando novamente respeito a T:

$$\begin{aligned} \frac{d^2\xi}{dT^2} = & \frac{(P_1 - P_o) \frac{d}{dT} \left\{ \left( \frac{dP}{dT} - \frac{dP_o}{dT} \right) - \frac{P - P_o}{P_1 - P_o} \left( \frac{dP_1}{dT} - \frac{dP_o}{dT} \right) \right\} -}{(P_1 - P_o)^2} \\ & - \frac{\left\{ \left( \frac{dP}{dT} - \frac{dP_o}{dT} \right) - \frac{P - P_o}{P_1 - P_o} \left( \frac{dP_1}{dT} - \frac{dP_o}{dT} \right) \right\} \left( \frac{dP_1}{dT} - \frac{dP_o}{dT} \right)}{(P_1 - P_o)^2} \end{aligned} \quad (A.2.2)$$

Tomando separadamente a derivada do parêntesis :

$$\frac{d}{dT} \left\{ \left( \frac{dP}{dT} - \frac{dP_o}{dT} \right) - \frac{P - P_o}{P_1 - P_o} \left( \frac{dP_1}{dT} - \frac{dP_o}{dT} \right) \right\} = \frac{dA}{dT} - \frac{dB}{dT} \quad (A.2.3)$$

Onde o termo A é dado por:

$$A = \left( \frac{dP}{dT} - \frac{dP_o}{dT} \right) \quad e \quad B \text{ por } B = \frac{P - P_o}{P_1 - P_o} \left( \frac{dP_1}{dT} - \frac{dP_o}{dT} \right)$$

$$\text{Então} \quad \frac{dA}{dT} = \frac{d^2P}{dT^2} - \frac{d^2P_o}{dT^2} \quad e \quad (A.2.4)$$

$$\begin{aligned} \frac{dB}{dT} = & \frac{d}{dT} \left[ \frac{P - P_o}{P_1 - P_o} \left( \frac{dP_1}{dT} - \frac{dP_o}{dT} \right) \right] = \\ & \frac{\left\{ \left( \frac{d^2P_1}{dT^2} - \frac{d^2P_o}{dT^2} \right) (P - P_o) + \left( \frac{dP_1}{dT} - \frac{dP_o}{dT} \right) \left( \frac{dP}{dT} - \frac{dP_o}{dT} \right) \right\} (P_1 - P_o) - \left( \frac{dP_1}{dT} - \frac{dP_o}{dT} \right) (P - P_o) \left( \frac{dP_1}{dT} - \frac{dP_o}{dT} \right)}{(P_1 - P_o)^2} \end{aligned} \quad (A.2.5)$$

Substituindo (A.2.5) e (A.2.4) em (A.2.3) e o resultado em (A.2.2) se obtém depois de simplificar:

$$\frac{d^2\xi}{dT^2} = \frac{1}{P_1 - P_o} \left\{ \frac{d^2P}{dT^2} - \frac{2}{P_1 - P_o} \left( \frac{dP_1}{dT} - \frac{dP_o}{dT} \right) \left( \frac{dP}{dT} - \frac{dP_o}{dT} \right) + \frac{2}{(P_1 - P_o)^2} \left( \frac{dP_1}{dT} - \frac{dP_o}{dT} \right)^2 (P - P_o) \right\} \quad (A.2.6)$$

### Anexo 3.

Retomando a equação que queremos transformar:

$$\left(\theta \frac{d^2\theta}{dT^2}\right)_{Ti} - [n(\theta^n - 1) + 1] \left(\frac{d\theta}{dT}\right)_{Ti}^2 = -C(T)\theta \left(\frac{d\theta}{dT}\right)_{Ti} \quad (A.3.1)$$

$$\text{Como: } \theta(T) = \frac{T^2 R}{\beta E} \left( Ko \exp\left(-\frac{E}{RT}\right) \right) \left(1 - 2\frac{RT}{E}\right) = \frac{T^2 RK}{\beta E} \left(1 - 2\frac{RT}{E}\right) \quad (A.3.2)$$

$$\frac{d\theta}{dT} = \frac{KoR}{\beta E} \frac{d}{dT} \left( e^{-\frac{E}{RT}} T^2 \left(1 - \frac{2RT}{E}\right) \right)$$

$$\begin{aligned} \frac{d}{dT} \left( e^{-\frac{E}{RT}} T^2 \left(1 - \frac{2RT}{E}\right) \right) &= \\ &= -\frac{2e^{-\frac{E}{RT}} RT^2}{E} + \frac{e^{-\frac{E}{RT}} E \left(1 - \frac{2RT}{E}\right)}{R} + 2e^{-\frac{E}{RT}} T \left(1 - \frac{2RT}{E}\right) \\ \frac{d\theta}{dT} &= \frac{RKo}{\beta E} \left[ -\frac{2e^{-\frac{E}{RT}} RT^2}{E} + \frac{e^{-\frac{E}{RT}} E \left(1 - \frac{2RT}{E}\right)}{R} + 2e^{-\frac{E}{RT}} T \left(1 - \frac{2RT}{E}\right) \right] \\ &= \frac{RK}{\beta E} \left[ -\frac{2RT^2}{E} + \frac{E \left(1 - \frac{2RT}{E}\right)}{R} + 2T \left(1 - \frac{2RT}{E}\right) \right] \end{aligned} \quad (A.3.3)$$

$$\frac{d^2\theta}{dT^2} = \frac{RKo}{\beta E} \left[ \frac{e^{-\frac{E}{RT}} (E^3 - 6ER^2T^2 - 12R^3T^3)}{ER^2T^2} \right] = \frac{RK}{\beta E} \frac{(E^3 - 6ER^2T^2 - 12R^3T^3)}{ER^2T^2} \quad (A.3.4)$$

Calculando o termino da equação:

$$n(\theta^n - 1) + 1$$

E Como já se conhece que:  $\theta^n \approx 1 + n \ln \theta + \frac{n^2 (\ln \theta)^2}{2}$ .

$$\begin{aligned} n(\theta^n - 1) + 1 &= n \left( 1 + n \ln \theta + \frac{n^2 (\ln \theta)^2}{2} - 1 \right) + 1 = n^2 \ln \theta + \frac{n^3 (\ln \theta)^2}{2} + 1 = \\ &= 1 + n^2 (\ln[bTx] + \ln[1 - 2x]) + \frac{1}{2} n^3 (\ln[bTx] + \ln[1 - 2x])^2 \end{aligned}$$

E como  $\ln(1 - 2x) \cong -2x$  Se  $x \ll 1$

$$\begin{aligned} n(\theta^n - 1) + 1 &= 1 + n^2 (\ln[bTx] + \ln[1 - 2x]) + \frac{1}{2} n^3 (\ln[bTx] + \ln[1 - 2x])^2 = 1 + \\ &= n^2 (\ln[bTx] - 2x) + \frac{1}{2} n^3 (\ln[bTx] - 2x)^2 \end{aligned} \quad (A.3.5)$$

Onde temos designado:  $x = \frac{RT}{E}$  e  $b = \frac{K}{\beta}$

Fazendo uso de esta representação as equações anteriores podem se escrever como:

$$\frac{d\theta}{dT} = b(1 - 6x^2) \quad (\text{A.3.3b})$$

$$\theta = \frac{T^2 RK}{\beta E} \left(1 - 2 \frac{RT}{E}\right) = bTx(1 - 2x) \quad (\text{A.3.2 b})$$

$$\frac{d^2\theta}{dT^2} = \frac{RK}{\beta E} \frac{(E^3 - 6ER^2T^2 - 12R^3T^3)}{ER^2T^2} = \frac{bR}{E} (x^{-2} - 12x - 6) \quad (\text{A.3.4 b})$$

$$\theta \frac{d^2\theta}{dT^2} = -2b^2x + b^2 - 6b^2x^2 + 24b^2x^4 \quad (\text{A.3.6})$$

Tinindo presente que  $x \ll 1$  (em o rango de trabalho) e por tanto depreciando as potencias superiores de  $x$  e substituindo (A.3.5), (A.3.3b), (A.3.2 b) e (A.3.4 b) ou (A.3.6) em a equação (A.3.1), se obtém:

$$b^2(1 - 2x) - (n^2 \ln \theta + \frac{n^3}{2} (\ln \theta)^2 + 1)b^2 = -C(T)b^2Tx$$

Simplificando e substituindo (A.3.5) em a ultima equação:

$$2x + n^2(\ln(bTx) - 2x) + \frac{n^3}{2}(\ln(bTx) - 2x)^2 = c(T)Tx \quad (\text{A.3.7})$$

E esta ultima expressão a podemos escrever como:

$$\ln(bTx) - 2x + \frac{n}{2} (\ln^2(bTx) - 4x \ln(bTx) + 4x^2) = \frac{CTx}{n^2} - \frac{2x}{n^2}$$

E desprezado a  $4x^2$

$$\ln(bTx) - 2x + \frac{n}{2} (\ln^2(bTx) - 4x \ln(bTx)) = \frac{CTx}{n^2} - \frac{2x}{n^2}$$

$$\ln(bTx) = \frac{CTx}{n^2} - \frac{2x}{n^2} + 2x - \frac{n}{2} (\ln^2(bTx) - 4x \ln(bTx))$$

$$\ln(bTx) = \frac{CTx}{n^2} + 2x \left(1 - \frac{1}{n^2}\right) + \frac{n}{2} (4x \ln(bTx) - \ln^2(bTx))$$

$$\ln(bTx) = \frac{CTx}{n^2} + 2x \left(1 - \frac{1}{n^2}\right) + 2nx \left(\ln(bTx) - \frac{\ln^2(bTx)}{4x}\right)$$

E para os valores típicos de trabalho o ultimo somando dentro do parêntesis é sempre menor que o primeiro e se pode depreciar.

Depois de substituir  $x = \frac{RT}{E}$  e  $b = \frac{K}{\beta}$ , se obtém a expressão final de trabalho.

$$\ln\left(\frac{Ti^2}{\beta}\right)_{Ti} = \ln\left(\frac{E}{RKi}\right) + \frac{E}{RTi} + \frac{C}{n^2} \frac{RTi^2}{E} + 2 \frac{RTi}{E} \left\{1 - \frac{1}{n^2} + n \ln\left[\frac{RKTi^2}{\beta E}\right]\right\}$$





## Anexo 4

Se adotarmos o formalismo de KJMA para a função transformada; a seguinte justificação do método de Kissinger pode ser dada:

$$\xi = 1 - \exp(-\theta^n) \quad (\text{A.4.1})$$

Por isso:

$$\frac{d\xi}{dt} = \frac{d\xi}{d\theta} \frac{d\theta}{dt} \quad (\text{A.4.2})$$

Como é conhecida, a segunda derivada da função transformada no ponto de inflexão é nula:

$$\left(\frac{d^2\xi}{dt^2}\right)_{Tp} = \frac{d}{dt} \left(\frac{d\xi}{d\theta} \frac{d\theta}{dt}\right) = \frac{d\theta}{dt} \frac{d}{d\theta} \frac{d\xi}{d\theta} + \frac{d\xi}{d\theta} \frac{d^2\theta}{dt^2} = \frac{d\theta}{dt} \frac{d}{d\theta} \left(\frac{d\xi}{d\theta}\right) \frac{d\theta}{dt} + \frac{d\xi}{d\theta} \frac{d^2\theta}{dt^2} = 0 \quad (\text{A.4.3})$$

A equação anterior pode ser resumida:

$$\left(\frac{d^2\xi}{dt^2}\right)_{Tp} = \left(\frac{d^2\xi}{d\theta^2} \left(\frac{d\theta}{dt}\right)^2\right)_{Tp} + \left(\frac{d\xi}{d\theta} \frac{d^2\theta}{dt^2}\right)_{Tp} = 0 \quad (\text{A.4.4})$$

Retomando a equação (5) :

$$d\theta = K(T) dt = K_0 \exp\left(-\frac{E}{R^*T}\right) dt \quad (\text{A.4.5})$$

É evidente de aqui que :2

$$\left(\frac{d\theta}{dt}\right)^2 = K(T)^2 \quad (\text{A.4.6})$$

$$\frac{d^2\theta}{dt^2} = K(T) \frac{E}{RT^2} \frac{dT}{dt} \quad (\text{A.4.7})$$

Dividindo (A.4.6) por (A.4.7):

$$\left(\frac{d\theta}{dt}\right)^2 \bigg/ \frac{d^2\theta}{dt^2} = \frac{KRT^2}{E\beta} \approx \theta \quad (\text{A.4.8})$$

$$\text{Já que: } \theta(T) = \frac{T^2 R}{\beta E} \left(K_0 \exp\left(-\frac{E}{RT}\right)\right) \left(1 - 2 \frac{RT}{E}\right) = K(T) \frac{T^2 R}{\beta E} \left(1 - 2 \frac{RT}{E}\right) \approx \frac{KRT^2}{E\beta} \quad (\text{A.1.11})$$

Se  $RT/E \ll 1$

Inserindo a equação (A.4.8) na equação (A.4.4):

$$\left(\frac{d^2\xi}{dt^2}\right)_{Tp} = \left(\theta \frac{d^2\theta}{dt^2} \frac{d^2\xi}{d\theta^2} + \frac{d\xi}{d\theta} \frac{d^2\theta}{dt^2}\right)_{Tp} = 0$$

De donde finalmente obtém-se que:

$$\left(\theta \frac{d^2\xi}{d\theta^2} + \frac{d\xi}{d\theta}\right)_{Tp} = 0 \quad (\text{A.4.9})$$

**A solução de esta equação é  $\vartheta p=1$ .** Por isso a temperatura  $T_p$  onde a taxa da reação é máxima, ou seja, no ponto de inflexão da curva  $\xi$  VS  $T$  ( ou  $t$  ) ,ocorre sempre com boa aproximação, para o mesmo valor de  $\xi$  para diferentes taxas de aquecimento  $\beta = \frac{dT}{dt}$  é por isto é razoável substituir  $T_i$  ( valor do ponto de inflexão nas curvas dilatométricas  $P$  vs  $T$  (ou  $t$ )) por  $T_p$  ,o ponto de inflexão da curva  $\xi$  VS  $T$  ( ou  $t$  ) [42].

Este mesmo resultado foi obtido por T. Ozawa (1984) [79] assumindo que a fração transformada segue uma distribuição de Poisson. Também foi obtido por Farjas J, Roura P (2006) [55] ver a equações (51) e (52).

Estes razoamentos constituem uma justificativa para o método de Kissinger de determinação da energia de ativação. Em efeito sim tomamos a expressão (A.1.11) obtida diretamente da integral de temperatura , sem recorrer a nenhum modelo cinético:

$$\theta_f = K(T) \frac{T_f^2 R}{\beta E} \left( 1 - 2 \frac{RT_f}{E} \right)$$

E tomamos um estado fixo da transformação, digamos  $f$ ,  $T_f$ . Como normalmente  $RT_f/E \ll 1$ :

$$\theta_f \approx K(T) \frac{T_f^2 R}{\beta E} = K \exp\left(-\frac{E}{RT}\right) \frac{T_f^2 R}{\beta E} \quad (\text{A.4.10})$$

Aplicando logaritmo natural a (A.4.10) e agrupando convenientemente:

$$\ln \frac{T_f^2}{\beta} = \frac{E}{RT_f} + \ln \frac{E}{RK_0} + \ln \theta_f \quad (\text{A.4.10})$$

O resultado (A.4.10) será usado para determinar  $E=E(f)$ . Note que para o ponto de inflexão  $T=T_p$ ,  $\theta_{f(T_p)}=1$  o logaritmo de  $\theta_{f(T_p)}$  se anula .Esta é uma equação do tipo de Kissinger. Que pode ser usada também para calcular  $K_0$ .



## Anexo 5 Cálculo da Energia de Ativação por Regressão Linear

```

R = 0.00831;
TgKelvin[x_] := 273.15 + x;
φ1 = 5;
φ2 = 10;
φ3 = 15;
φ4 = 20;
φ5 = 30;
T1 = 138.5;
T2 = 145.8;
T3 = 151.2;
T4 = 155.35;
T5 = 159.3;
rapienfria = {φ1, φ2, φ3, φ4, φ5};
listatepgradCent = {T1, T2, T3, T4, T5};
TemperaturaKelvin = {};
Do[ti = listatepgradCent[[i]];
  TemperaturaKelvin = AppendTo[TemperaturaKelvin, TgKelvin[ti]],
  {i, 1, Length[listatepgradCent]}]
TemperaturaKelvin
listay = {};
Do[Te = TemperaturaKelvin[[i]]; flujo = rapienfria[[i]];
  listay = AppendTo[listay, {Te^-1, Log[Te^2 / flujo]}],
  {i, 1, Length[TemperaturaKelvin]}]
listay
result = NonlinearModelFit[listay, a + b * x, {a, b}, {x}]
l = Normal[result]
Print["Pendiente =", " ", pend = l[[2, 1]]]
Print["intercepto =", " ", Interc = l[[1]]]
Print["Energiaactiva=", " ", Energiaactiva1 = pend * R]
Print["Ko =", " ",  $\frac{\text{Energiaactiva1} * \text{Exp}[-\text{Interc}]}{R}$ ]
nlm = NonlinearModelFit[listay, a + b x, {a, b}, {x}] [
  {"ANOVATable", "ParameterPValues", "AdjustedRSquared", "RSquared",
    "ParameterTStatistics", "ParameterConfidenceIntervals",
    "ParameterConfidenceIntervalTable"}]
Print["valor de energia superior =", R * nlm[[6, 2, 2]]]
Print["valor de energia inferior =", R * nlm[[6, 2, 1]]]

Show[ListPlot[listay],
  Plot[result[x], {x, 0.0023050803004043903, 0.002433}],
  Frame → True, AxesLabel →  $\left\{\frac{1}{T_i}, \frac{\ln T_i^2}{\phi}\right\}$ ]

Plot[result[x], {x, 0.0024292481476982874, 0.0023124060585038735},
  Epilog → Point[listay], PlotStyle → {Red, Thick},
  Frame → True, AxesLabel →  $\left\{\frac{1}{T_i}, \frac{\ln T_i^2}{\phi}\right\}$ ]
NonlinearModelFit[listay, a + b x, {a, b}, {x}] [
  {"ANOVATable", "ParameterPValues", "AdjustedRSquared", "RSquared",

```

```

"ParameterTStatistics", "ParameterConfidenceIntervalTable"]
lm = LinearModelFit[listay, x, x]
listasalida = {};
listasalida = lm[{"AdjustedRSquared", "RSquared", "FitResiduals"}]
 $\sqrt{\text{listasalida}[[2]]}$ 
Listares = {};
Listares = listasalida[[3]]
ListPlot[Listares, PlotStyle → {Hue[0.5], PointSize[0.04]}]
listasalida2 = {};
listasalida2 =
  lm[{"FitResiduals", "SinglePredictionConfidenceIntervalTable",
    "ParameterConfidenceRegion"}];
errors = listasalida2[[1]]
tablasimpledeintecnf = listasalida2[[2]]
Observed = {}
Do[Observed = AppendTo[Observed, tablasimpledeintecnf[[1, 1, i, 1]]],
  {i, 2, 6}]
Observed
predicted = {};
Do[predicted = AppendTo[predicted, tablasimpledeintecnf[[1, 1, i, 2]]],
  {i, 2, 6}]
predicted
ListPlot[Transpose[{predicted, errors}], PlotStyle → PointSize[0.02`]]
gplaj = Plot[lm[x], {x, 0.0024292481476982874, 0.0023124060585038735}];
ci = {}
Do[ci = AppendTo[ci, tablasimpledeintecnf[[1, 1, i, 4]]], {i, 2, 6}]
ci
(xval = First /@ listay; predicted = Transpose[{xval, predicted}];
  lowerCI = Transpose[{xval, First /@ ci}];
  upperCI = Transpose[{xval, Last /@ ci}]);
gfrs = ListPlot[{listay, predicted, lowerCI, upperCI},
  Joined → {False, True, True, True}, PlotStyle → {Automatic,
    Automatic, Dashing[{0.05`, 0.05`}], Dashing[{0.05`, 0.05`}]}}]
Show[gfrs, gplaj]

```

Out[17]= {411.65, 418.95, 424.35, 428.5, 432.45}

Out[20]= {{0.00242925, 10.4309}, {0.00238692, 9.77292},  
 {0.00235655, 9.39307}, {0.00233372, 9.12485}, {0.00231241, 8.73774}}

Out[21]= FittedModel[ $-23.7799 + 14075.7 x$ ]

Out[22]=  $-23.7799 + 14075.7 x$

Pendiente = 14075.7

intercepto = -23.7799

Energiaactiva= 116.969

Ko =  $2.99184 \times 10^{14}$

Out[25]= Null<sup>2</sup>

Out[26]=

	DF	SS	MS
Model	2	452.148	226.074
Error	3	0.00641532	0.00213844
Uncorrected Total	5	452.154	
Corrected Total	4	1.67397	

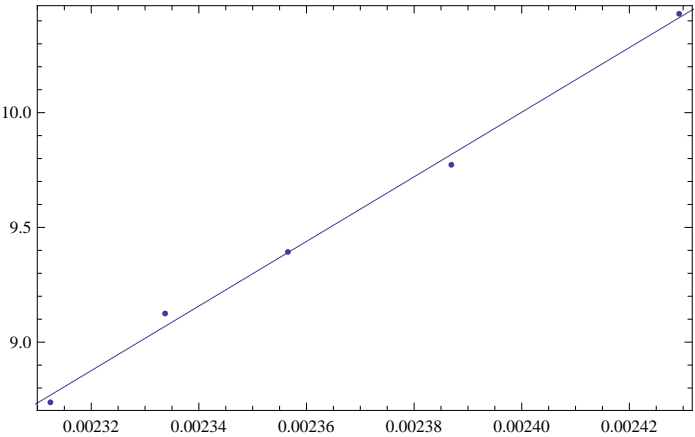
{0.00027503, 0.000100808}, 0.999976, 0.999986,  
{-19.9553, 27.9249}, {{-27.5722, -19.9875}, {12471.6, 15679.9}},

	Estimate	Standard Error	Confidence Interval
a	-23.7799	1.19165	{-27.5722, -19.9875}
b	14075.7	504.057	{12471.6, 15679.9}

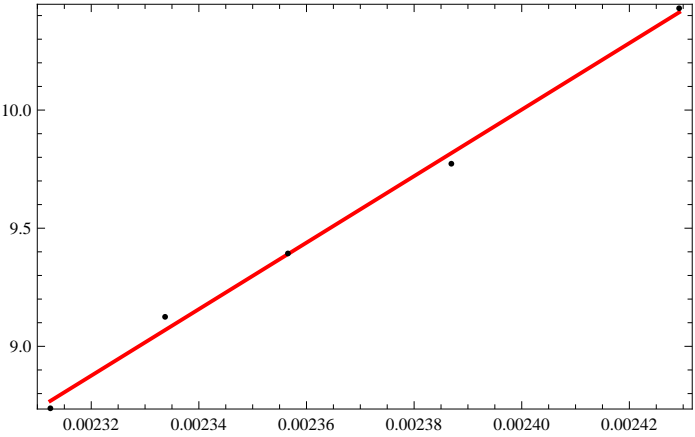
valor de energia superior =130.3

valor de energia inferior =103.639

Out[29]=



Out[30]=



Out[31]=

	DF	SS	MS
Model	2	452.148	226.074
Error	3	0.00641532	0.00213844
Uncorrected Total	5	452.154	
Corrected Total	4	1.67397	

{0.00027503, 0.000100808}, 0.999976, 0.999986, {-19.9553, 27.9249},

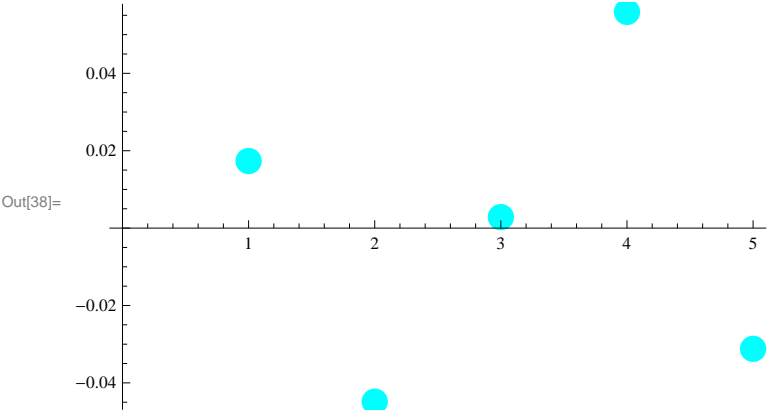
	Estimate	Standard Error	Confidence Interval
a	-23.7799	1.19165	{-27.5722, -19.9875}
b	14 075.7	504.057	{12 471.6, 15 679.9}

Out[32]= FittedModel[ -23.7799 + 14 075.7 x ]

Out[34]= {0.99489, 0.996168, {0.0173364, -0.0448505, 0.00283999, 0.0558729, -0.0311988}}

Out[35]= 0.998082

Out[37]= {0.0173364, -0.0448505, 0.00283999, 0.0558729, -0.0311988}



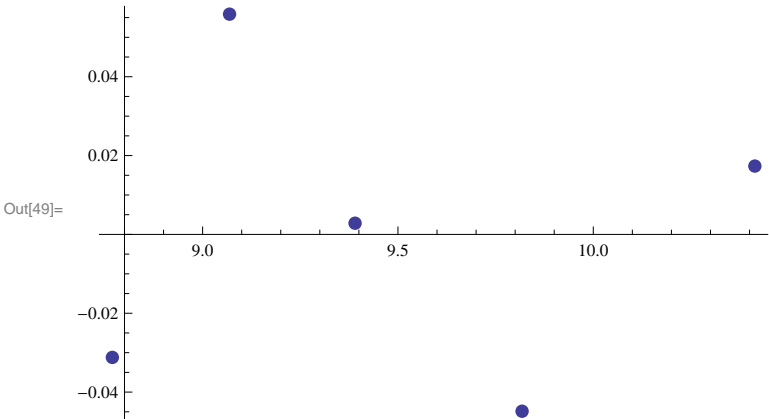
Out[41]= {0.0173364, -0.0448505, 0.00283999, 0.0558729, -0.0311988}

	Observed	Predicted	Standard Error	Confidence Interval
	10.4309	10.4136	0.0604607	{10.2212, 10.606}
Out[42]=	9.77292	9.81777	0.0519837	{9.65233, 9.9832}
	9.39307	9.39023	0.0507876	{9.2286, 9.55186}
	9.12485	9.06898	0.0528725	{8.90071, 9.23724}
	8.73774	8.76893	0.0568893	{8.58789, 8.94998}

Out[43]= { }

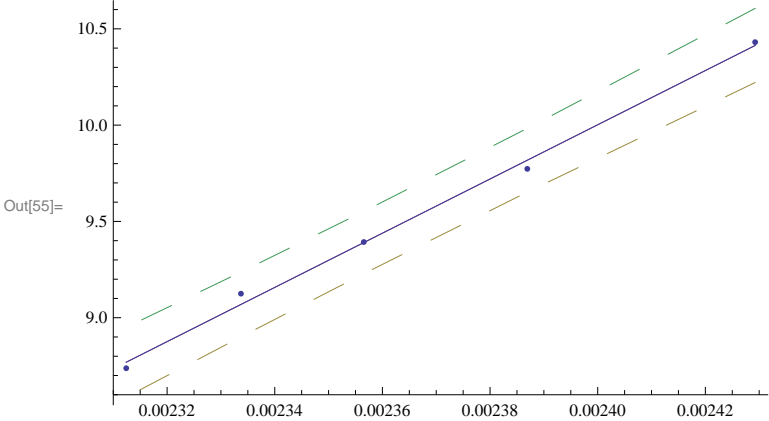
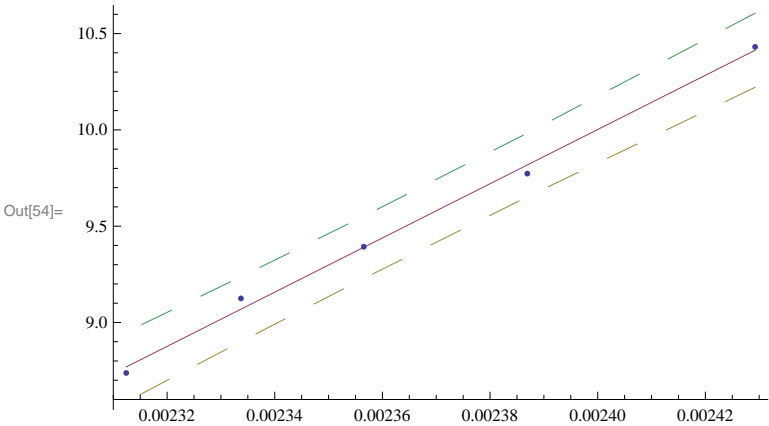
Out[45]= {10.4309, 9.77292, 9.39307, 9.12485, 8.73774}

Out[48]= {10.4136, 9.81777, 9.39023, 9.06898, 8.76893}



Out[51]= { }

Out[53]= {{10.2212, 10.606}, {9.65233, 9.9832},  
{9.2286, 9.55186}, {8.90071, 9.23724}, {8.58789, 8.94998}}





(\* Anexo 5 b\*) Cálculo da Energia de Ativação por Regressão Linear, Processo II

```

R = 0.00831;
TgKelvin[x_] := 273.15 + x;
φ1 = 5;
φ2 = 10;
φ3 = 15;
φ4 = 20;
φ5 = 30;
T1 = 311.2` ;
T2 = 320.4` ;
T3 = 326;
T4 = 330.` ;
T5 = 335.7` ;
rapiefria = {φ1, φ2, φ3, φ4, φ5};
listatepgradCent = {T1, T2, T3, T4, T5};
TemperaturaKelvin = {};
Do[ti = listatepgradCent[[i]];
  TemperaturaKelvin = AppendTo[TemperaturaKelvin, TgKelvin[ti]],
  {i, 1, Length[listatepgradCent]}]
TemperaturaKelvin
listay = {};
Do[Te = TemperaturaKelvin[[i]]; flujo = rapiefria[[i]];
  listay = AppendTo[listay, {Te^-1, Log[Te^2 / flujo]}],
  {i, 1, Length[TemperaturaKelvin]}]
listay
result = NonlinearModelFit[listay, a + b * x, {a, b}, {x}]
l = Normal[result]
Print["Pendiente =", " ", pend = l[[2, 1]]]
Print["intercepto =", " ", Interc = l[[1]]]
Print["Energiaactiva=", " ", Energiaactiva1 = pend * R]
Print["Ko =", " ",  $\frac{\text{Energiaactiva1} * \text{Exp}[-\text{Interc}]}{R}$ ]
nlm = NonlinearModelFit[listay, a + b x, {a, b}, {x}] [
  {"ANOVATable", "ParameterPValues", "AdjustedRSquared", "RSquared",
    "ParameterTStatistics", "ParameterConfidenceIntervals",
    "ParameterConfidenceIntervalTable"}]
Print["valor de energia superior =", R * nlm[[6, 2, 2]]]
Print["valor de energia inferior =", R * nlm[[6, 2, 1]]]

Show[ListPlot[listay],
  Plot[result[x], {x, 0.0017113031573543255`, 0.001642440666830911`}],
  Frame → True, AxesLabel →  $\left\{ \frac{1}{T_i}, \frac{\ln T_i^2}{\phi} \right\}$ ]

Plot[result[x], {x, 0.0017113031573543255`, 0.001642440666830911`},
  Epilog → Point[listay], PlotStyle → {Red, Thick},
  Frame → True, AxesLabel →  $\left\{ \frac{1}{T_i}, \frac{\ln T_i^2}{\phi} \right\}$ ]

NonlinearModelFit[listay, a + b x, {a, b}, {x}] [
  {"ANOVATable", "ParameterPValues", "AdjustedRSquared", "RSquared",

```

```

    "ParameterTStatistics", "ParameterConfidenceIntervalTable"]
lm = LinearModelFit[listay, x, x]
listasalida = {};
listasalida = lm[{"AdjustedRSquared", "RSquared", "FitResiduals"}]
 $\sqrt{\text{listasalida}[[2]]}$ 
Listares = {};
Listares = listasalida[[3]]
ListPlot[Listares, PlotStyle → {Hue[0.5], PointSize[0.04]}]
listasalida2 = {};
listasalida2 =
    lm[{"FitResiduals", "SinglePredictionConfidenceIntervalTable",
        "ParameterConfidenceRegion"}];
errors = listasalida2[[1]]
tablasimpledeintecnf = listasalida2[[2]]
Observed = {}
Do[Observed = AppendTo[Observed, tablasimpledeintecnf[[1, 1, i, 1]]],
    {i, 2, 6}]
Observed
predicted = {};
Do[predicted = AppendTo[predicted, tablasimpledeintecnf[[1, 1, i, 2]]],
    {i, 2, 6}]
predicted
ListPlot[Transpose[{predicted, errors}], PlotStyle → PointSize[0.02`]]
gplaj = Plot[lm[x], {x, 0.0017113031573543255`, 0.001642440666830911`}];
ci = {}
Do[ci = AppendTo[ci, tablasimpledeintecnf[[1, 1, i, 4]]], {i, 2, 6}]
ci
(xval = First/@listay; predicted = Transpose[{xval, predicted}];
    lowerCI = Transpose[{xval, First/@ci}];
    upperCI = Transpose[{xval, Last/@ci}]);
gfrs = ListPlot[{listay, predicted, lowerCI, upperCI},
    Joined → {False, True, True, True}, PlotStyle → {Automatic,
        Automatic, Dashing[{0.05`, 0.05`}], Dashing[{0.05`, 0.05`}]}}]
Show[gfrs, gplaj]

```

Out[127]= {584.35, 593.55, 599.15, 603.15, 608.85}

Out[130]= {{0.0017113, 11.1316}, {0.00168478, 10.4697},  
 {0.00166903, 10.083}, {0.00165796, 9.8086}, {0.00164244, 9.42195}}

Out[131]= FittedModel[ $-31.3166 + 24804. x$ ]

Out[132]=  $-31.3166 + 24804. x$

Pendiente = 24804.

intercepto = -31.3166

Energiaactiva= 206.121

Ko =  $9.8892 \times 10^{17}$

Out[135]= Null<sup>2</sup>

Out[136]=

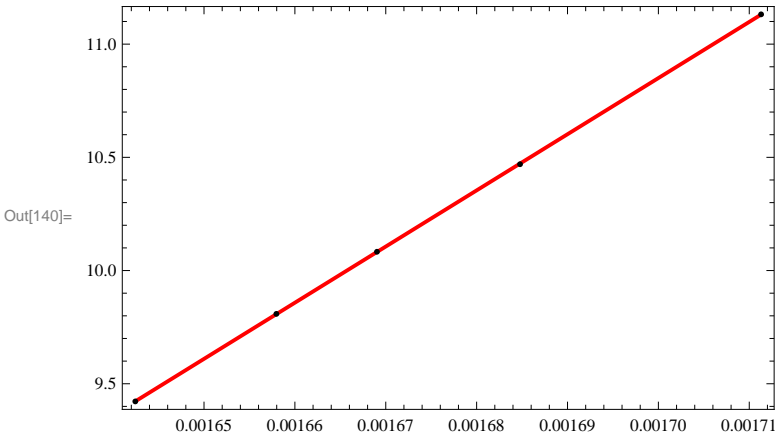
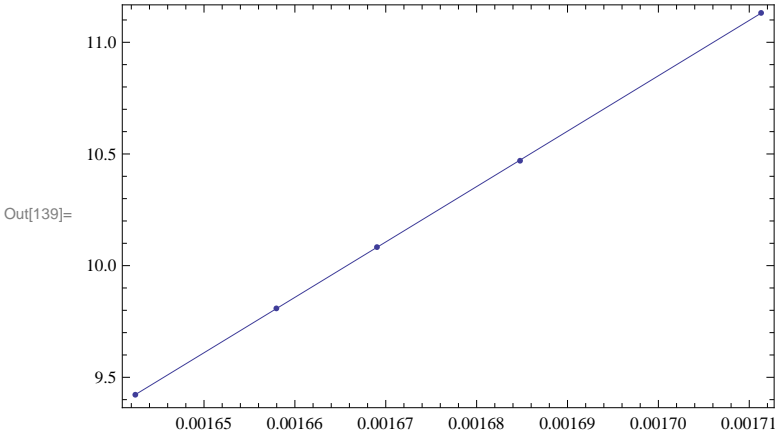
	DF	SS	MS
Model	2	520.173	260.087
Error	3	0.0000121801	$4.06004 \times 10^{-6}$
Uncorrected Total	5	520.173	
Corrected Total	4	1.71133	

$\{1.87586 \times 10^{-8}, 8.05874 \times 10^{-9}\}, 1., 1., \{-489.878, 649.232\},$   
 $\{-31.5201, -31.1132\}, \{24\,682.4, 24\,925.5\},$

	Estimate	Standard Error	Confidence Interval
a	-31.3166	0.0639274	$\{-31.5201, -31.1132\}$
b	24\,804.	38.2051	$\{24\,682.4, 24\,925.5\}$

valor de energia superior =207.131

valor de energia inferior =205.111



Out[141]=

	DF	SS	MS
Model	2	520.173	260.087
Error	3	0.0000121801	$4.06004 \times 10^{-6}$
Uncorrected Total	5	520.173	
Corrected Total	4	1.71133	

$\{1.87586 \times 10^{-8}, 8.05874 \times 10^{-9}\}, 1., 1., \{-489.878, 649.232\},$

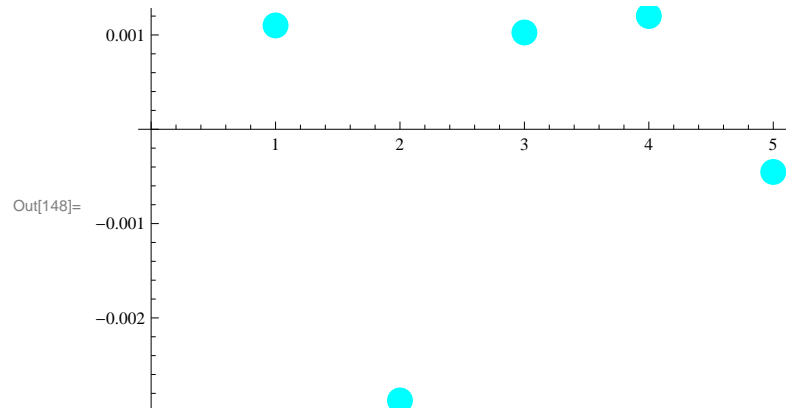
	Estimate	Standard Error	Confidence Interval
a	-31.3166	0.0639274	$\{-31.5201, -31.1132\}$
b	24804.	38.2051	$\{24682.4, 24925.5\}$

Out[142]= FittedModel [ -31.3166 + 24804. x ]

Out[144]= {0.999991, 0.999993, {0.00110094, -0.00287542, 0.00102605, 0.00120099, -0.000452557}}

Out[145]= 0.999996

Out[147]= {0.00110094, -0.00287542, 0.00102605, 0.00120099, -0.000452557}



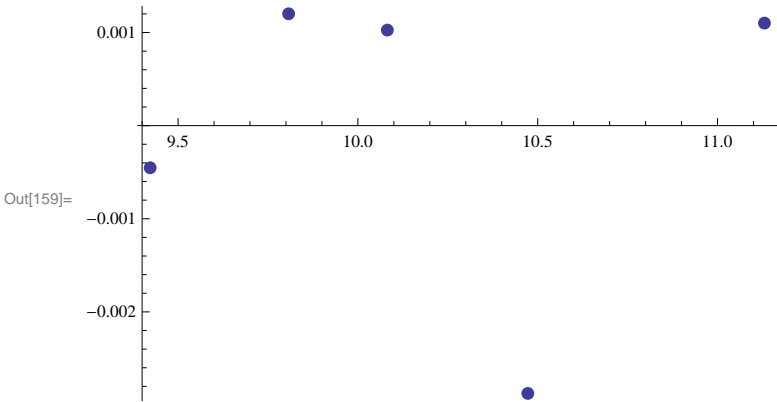
Out[151]= {0.00110094, -0.00287542, 0.00102605, 0.00120099, -0.000452557}

	Observed	Predicted	Standard Error	Confidence Interval
	11.1316	11.1305	0.00264613	{11.122, 11.1389}
Out[152]=	10.4697	10.4725	0.00225189	{10.4654, 10.4797}
	10.083	10.0819	0.00221275	{10.0749, 10.089}
	9.8086	9.8074	0.00228181	{9.80014, 9.81466}
	9.42195	9.4224	0.00249887	{9.41445, 9.43035}

Out[153]= { }

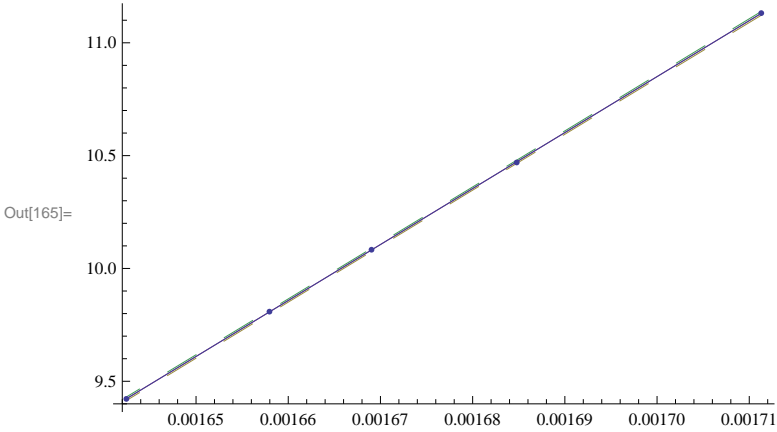
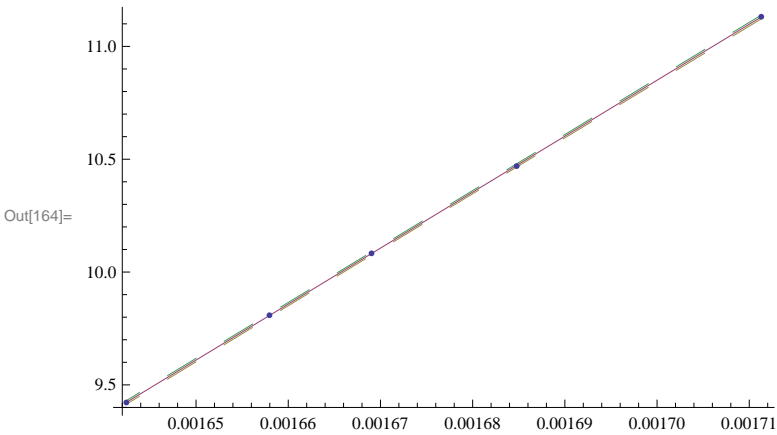
Out[155]= {11.1316, 10.4697, 10.083, 9.8086, 9.42195}

Out[158]= {11.1305, 10.4725, 10.0819, 9.8074, 9.4224}



Out[161]= { }

Out[163]= {{11.122, 11.1389}, {10.4654, 10.4797},  
{10.0749, 10.089}, {9.80014, 9.81466}, {9.41445, 9.43035}}



■ (\* Anexo 6 ,tuda a equação e com parametros \*)

```

R = 0.00831;
TgKelvin[x_] := 273.15 + x;
β1 = 5;
β2 = 10;
β3 = 15;
β4 = 20;
β5 = 30;
T1 = 138.5;
T2 = 145.8;
T3 = 151.2;
T4 = 155.35;
T5 = 159.3;
rapiefria = {β1, β2, β3, β4, β5};
listatepgradCent = {T1, T2, T3, T4, T5};
TemperaturaKelvin = {};
Do[ti = listatepgradCent[[i]];
  TemperaturaKelvin = AppendTo[TemperaturaKelvin, TgKelvin[ti]],
  {i, 1, Length[listatepgradCent]}]
TemperaturaKelvin
listay = {};
Do[Te = TemperaturaKelvin[[i]]; flujo = rapiefria[[i]];
  listay = AppendTo[listay, {Te^-1, Log[Te^2 / flujo]}],
  {i, 1, Length[TemperaturaKelvin]}]
listay
resul = NonlinearModelFit[listay,
  
$$\frac{\text{Ener}}{R} * y + \text{Log}\left[\frac{\text{Ener}}{R * Ko}\right], \{\{\text{Ener}, 115\}, \{Ko, 1 * 10^{14}\}\}, \{y\}]
resul[[1, 2]]
resul[[1, 2, 1, 2]]
resul[[1, 2, 2, 2]]
listasalida = {};
listasalida =
  resul[{"BestFitParameters", "FitResiduals", "ParameterConfidenceIntervals",
    "ParameterConfidenceIntervalTable", "ParameterTable"}]
residuos = listasalida[[2]]
valoresdex = First /@ listay
Lsitgraresiduos = {};
Do[Lsitgraresiduos = AppendTo[Lsitgraresiduos, {valoresdex[[i]], residuos[[i]]}],
  {i, 1, Length[residuos]}]
ListPlot[Lsitgraresiduos, AxesLabel → {Xexp, Residuos},
  PlotStyle → {Hue[0.7], PointSize[0.02]}]

{411.65, 418.95, 424.35, 428.5, 432.45}

{{0.00242925, 10.4309}, {0.00238692, 9.77292},
  {0.00235655, 9.39307}, {0.00233372, 9.12485}, {0.00231241, 8.73774}}

FittedModel[
$$-23.7799 + 14075.7 y$$
]

{Ener → 116.969, Ko → 2.99184 × 1014}
116.969$$

```

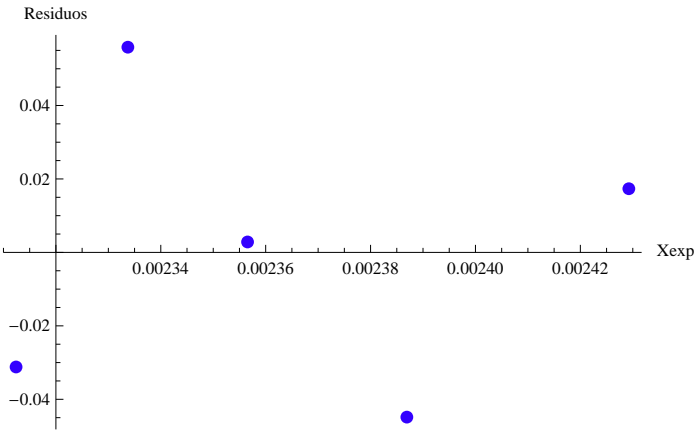
$2.99184 \times 10^{14}$

$\{ \{ \text{Ener} \rightarrow 116.969, \text{Ko} \rightarrow 2.99184 \times 10^{14} \},$   
 $\{ 0.0173364, -0.0448505, 0.00283999, 0.0558729, -0.0311988 \},$   
 $\{ \{ 116.745, 117.194 \}, \{ 2.99184 \times 10^{14}, 2.99184 \times 10^{14} \} \},$

	Estimate	Standard Error	Confidence Interval
Ener	116.969	0.0705728	{116.745, 117.194}
Ko	$2.99184 \times 10^{14}$	$8.04844 \times 10^{-16}$	$\{ 2.99184 \times 10^{14}, 2.99184 \times 10^{14} \}$

	Estimate	Standard Error	t Statistic	P-Value
Ener	116.969	0.0705728	1657.43	$4.84358 \times 10^{-10}$
Ko	$2.99184 \times 10^{14}$	$8.04844 \times 10^{-16}$	$3.71729 \times 10^{29}$	$4.2933 \times 10^{-89}$

$\{ 0.0173364, -0.0448505, 0.00283999, 0.0558729, -0.0311988 \}$   
 $\{ 0.00242925, 0.00238692, 0.00235655, 0.00233372, 0.00231241 \}$



## Anexo 7 Calculo dos Residuos

Para calcular os valores dos residuais Res1 e Res2 em a equação (13) é necessário calcular o valor do parâmetro Q, em neste caso o parâmetro P é a longitude da mostra, l:

$$Q(T_i) = 2 \frac{\frac{dp_1}{dT} - \frac{dp_o}{dT}}{p_1 - p_o} = 2 \frac{(\alpha_1 l_1 - \alpha_0 l_0)}{l_1 - l_0} \cong \frac{\Delta \alpha}{\frac{\Delta l}{l_o}} \quad (7.1)$$

Onde  $\alpha_0$  e  $\alpha_1$  são os coeficientes da dilatação lineal da amostra antes e depois da transformação, observar a figura 2.1. Para estruturas poliméricas tem sido verificada a seguinte relação [99]

$$\alpha = \alpha_1 V_1 + \alpha_2 V_2 + \alpha_3 V_3 + \dots \quad (7.2)$$

Onde  $\alpha_i$  e  $V_i$  são respectivamente o coeficiente de dilatação lineal e a fração de volume da fase i-énésima. Antes de iniciar a transformação veja a figura 3.2, segmento AO a equação (7.2) pode-se escrever da seguinte forma:

$$\alpha_o = \alpha_m V_m + \alpha_{ar} V_{ar} \quad (7.3)$$

E depois da transformação, processo I, veja novamente a figura 3.2:

$$\alpha_1 = \alpha_e V_e + \alpha_{fe} V_{fe} + \alpha_{ar} V_{ar} \quad (7.4)$$

Onde os sub índices (m,ar,e,fe), refreasse ao Martensita(m), Austenita residual (ar), carbeto Épsilon(e) e Ferrita(Fe) respectivamente. É certo que :

$$V_m + V_{ar} = 1, \quad V_e + V_{fe} + V_{ar} = 1 \quad (7.5)$$

Por isso em a equação (7.1) :

$$\Delta \alpha = \alpha_1 - \alpha_0 \quad (7.6)$$

É muito Bom conhecido que os câmbios relativos do volumem referidos ao volume inicial da amostra ficam relacionados com os câmbios relativos da longitude por a simples relação

$$\delta V/V = 3 \delta l/l.$$



Para transformações isomorfas. Pero não se comete um grão erro se aplicamos ela em este caso [82]. O cambio de volume se deve a que em o processo I o cambio das fases esta dado por:

$$(\alpha'_{0.5} + \gamma_{r0.04} \rightarrow \alpha_{0.2} + \varepsilon + \gamma_{r0.04}) \text{ Muito perto da temperatura Ti (tabela 3.2)}$$

É possível calcular a fração de volume da fase que se tem formado (O carboneto Épsilon),  $V_\varepsilon$ , sim as seguintes considerações são assumidas: Nomear os termos  $N^e, N^{fe}, N^m$  como o numero de átomos de ferro em as fases Épsilon, Ferrita e Martensita respectivamente, E seja também  $v^e, v^{fe}, v^m$  o volume de um átomo de ferro em as fases Épsilon, Ferrita e Martensita respectivamente. Assume-se um revenido isotérmico, então a seguinte relação é valida:

$$\frac{\Delta V}{V} = 3 \frac{\Delta l}{l_o} = \frac{N^e v^e + N^{fe} v^{fe} + N^{ar} v^{ar} - (N^e v^m + N^{fe} v^m + N^{ar} v^{ar})}{N^m v^m} \quad (7.7)$$

Tendo presente a conservação de sítios:

$$N^m = N^{fe} + N^e, \quad 1 = \frac{N^{fe}}{N^m} + \frac{N^e}{N^m} = \frac{N^{fe}}{N^m} + y, \quad y = \frac{N^e}{N^m} \quad (7.8)$$

Onde  $y = \frac{N^e}{N^m}$  é a fração de átomos de ferro que tem sido transformada em carbonetos épsilon. e agora a relação (VII.7) pode-se escrever como :

$$\frac{\Delta V}{V} = 3 \frac{\Delta l}{l_o} = \frac{y(v^e - v^m) + (1 - y)(v^{fe} - v^m)}{v^m} \quad (7.9)$$

A fração de volume que se tem formado de carbeto Épsilon :

$$V_E = \frac{N^e v^e}{N^m v^m} = y \frac{v^e}{v^m} \quad (VII.10)$$

Se se tem presente (7.10) e pode derivar se que :

$$V_E = \frac{N^e v^e}{N^m v^m} = y \frac{v^e}{v^m} = \left\{ \frac{\left( 3 \frac{\Delta l}{l_o} + 1 \right) v^m - v^{fe}}{v^e - v^{fe}} \right\} \frac{v^e}{v^m} \quad (7.11)$$

Em a relação anterior  $\Delta l/l_o$  é obtido diretamente do registro do dilatômetro extrapolando ate o ponto de inflexão da curva, veja a figura 2.1, os valores de referencia  $P_1$  e  $P_o$  são obtidos de

esta forma como é indicado em a figura 2.1. Posteriormente se subtrai o valor P1 de Po. Os valores  $v^e$ ,  $v^{fe}$ ,  $v^m$ , assim como todos os  $\alpha_i$  são obtidos da literatura[82]

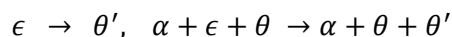
$$v^m = 11.897 \text{ A}^3/\text{Fe} \quad \alpha_m = 11.5 \cdot 10^{-6} \text{ C}^{-1}$$

$$v^{fe} = 11.77 \text{ A}^3/\text{Fe} \quad \alpha_{fe} = 14.5 \cdot 10^{-6} \text{ C}^{-1} \quad ; \quad \alpha_{ar} = 23 \cdot 10^{-6} \text{ C}^{-1};$$

$$v^e = 13.827 \text{ A}^3/\text{Fe} \quad \alpha_E = \alpha_\theta = 12.5 \cdot 10^{-6} \text{ C}^{-1}$$

Em a tabela seguinte se mostram os valores do parâmetro Q calculados para cada curva do registro dilatométrico para o primeiro processo. A fração de volume dos carbonetos Epsilon  $\epsilon$ ,  $V_\epsilon$ , se calcula usando a equação (7.11) e a fração de volume de Ferrita,  $V_{Fe}$  se calcula usando a relação (7.5). A fração de volume da martensita e da Austenita residual é em todos os casos:  $V_{ar} = 0.04$ ;  $V_m = 0.96$ .

Como poderá apreciar se o parâmetro Q é aproximadamente  $\approx -0.01$  para o primeiro processo do revenido. Para o segundo processo do revenido (terceiro estado do revenido) o parâmetro Q = 0. Pois a seguinte reação se desenvolve durante o segundo processo:



Onde  $\theta$  e  $\alpha$  são a Cementita e a Ferri ta respectivamente. A Cementita que se origina da decomposição dos carbonetos Épsilon se designa por  $\theta'$ . Então

$$\alpha_0 = V_{fe} \alpha_{fe} + \alpha_e V_e + \alpha_\theta V_\theta \quad \text{e} \quad \alpha_1 = V_{fe} \alpha_{fe} + V_\theta \alpha_\theta + \alpha_{\theta'} V_{\theta'}$$

De esta forma  $\Delta\alpha = \alpha_1 - \alpha_0 = \alpha_\theta V_\theta - V_e \alpha_e = 0$  devido a que  $V_e = V_\theta$  e  $\alpha_e = \alpha_\theta$  e por tanto Q = 0. Em resume :

Tabela A 7.1: Valores do parâmetro Q, para o primeiro processo.

$\beta$	$\Delta l/l_0 \times 10^{-4}$	$V_c(\%)$	$V_{fe}(\%)$	$\alpha_1 \times 10^{-6}$	$\alpha_0 \times 10^{-6}$	$Q \times 10^{-2}$
5	-4.2	9.8	86.2	14.5	11.8	-1.3
10	-3.86	9.9	86.1	14.5	11.8	-1.6
15	-4.52	9.7	86.3	14.5	11.8	-1.3
20	-4.4	9.8	86.2	14.5	11.8	-1.1
30	-4.5	9.7	86.3	14.5	11.8	-1.1

E finalmente os valores dos residuais se mostram em a tabela A VII.2

Tabela A7.2: Processo I (I): E=117KJ/mol ; n =1; Ko=2.99\*10<sup>14</sup> min<sup>-1</sup> ; Q=-0.01.

Processo (II): E=206 KJ/mol; Ko=9.5\*10<sup>17</sup>; n=0.66; Q= 0.0; Res1=0.0

Process ( I )				Process (II)	
β	T <sub>i</sub> , K	Res1	Res2	T <sub>i</sub> , K	Res2
5	411.6	0.12	0.00018	584.3	-0.061
10	418.9	0.124	-0.0035	593.5	-0.0627
15	424.3	0.128	-0.00067	599.1	-0.063
20	428.5	0.13	0.0028	603.1	-0.064
30	432.4	0.13	-0.0027	608.8	-0.064

$$RES_1 = \frac{Q}{n^2} \frac{RT_i^2}{E}$$

$$RES_2 = 2 \frac{RT_i}{E} * \left\{ 1 - \frac{1}{n^2} + n \ln \left[ \frac{Ti^2 RK(Ti)}{\beta E} \right] \right\}$$



## ■ (\* Anexo 8 Haciendo una estimacion del parametro cf de Farjas \*)

```

Clear[Ko, Cf, Ener];
 $\alpha_p = 0.632$ ;
Na = 6.02 * 10^23;
Kb = 1.381 * 10^-23;
R = 0.00831;
Ener = 116.969;
TgKelvin[x_] := 273.15 + x;
 $\beta_1 = 5$ ;
 $\beta_2 = 10$ ;
 $\beta_3 = 15$ ;
 $\beta_4 = 20$ ;
 $\beta_5 = 30$ ;
T1 = 138.5;
T2 = 145.8;
T3 = 151.2;
T4 = 155.35;
T5 = 159.3;
rapienfria = { $\beta_1$ ,  $\beta_2$ ,  $\beta_3$ ,  $\beta_4$ ,  $\beta_5$ };
listatepgradCent = {T1, T2, T3, T4, T5};
TemperaturaKelvin = {};
Do[ti = listatepgradCent[[i]];
  TemperaturaKelvin = AppendTo[TemperaturaKelvin, TgKelvin[ti]],
  {i, 1, Length[listatepgradCent]}]
TemperaturaKelvin
listay = {};
Do[Te = TemperaturaKelvin[[i]]; flujo = rapienfria[[i]];
  listay = AppendTo[listay, {Te^-1, Log[flujo / Te^2]}],
  {i, 1, Length[TemperaturaKelvin]}]
listay
resul = NonlinearModelFit[listay,
  
$$-\frac{\text{Ener}}{R} * y + \text{Log}\left[\frac{R * Ko * Cf}{\text{Ener}}\right], \{\{Ko, 2.99184 * 10^{14}\}, \{Cf, 1.01\}\}, \{y\}]$$

Listasali = {};
Listasali = resul[{"BestFitParameters", "ANOVATable", "EstimatedVariance",
  "FitResiduals", "ParameterTable", "ParameterConfidenceIntervals",
  "ParameterConfidenceIntervalTable", "RSquared", "AdjustedRSquared"}]
tablaAnov = resul["ANOVATable"];
valorSS = tablaAnov[[1, 1, 3, 3]];
valorST = tablaAnov[[1, 1, 5, 3]];

Print["valordeR=",  $\sqrt{1 - \frac{\text{valorSS}}{\text{valorST}}}$ ]

Listares = {};
Listares = Listasali[[4]]
ListPlot[Listares, PlotStyle -> {Hue[0.5], PointSize[0.04]]}

{411.65, 418.95, 424.35, 428.5, 432.45}

{{0.00242925, -10.4309}, {0.00238692, -9.77292},
  {0.00235655, -9.39307}, {0.00233372, -9.12485}, {0.00231241, -8.73774}}

```

FittedModel[23.7798- 14 075.7 y]

	DF	SS	MS
Model	2	452.148	226.074
Error	3	0.00641532	0.00213844
Uncorrected Total	5	452.154	
Corrected Total	4	1.67397	

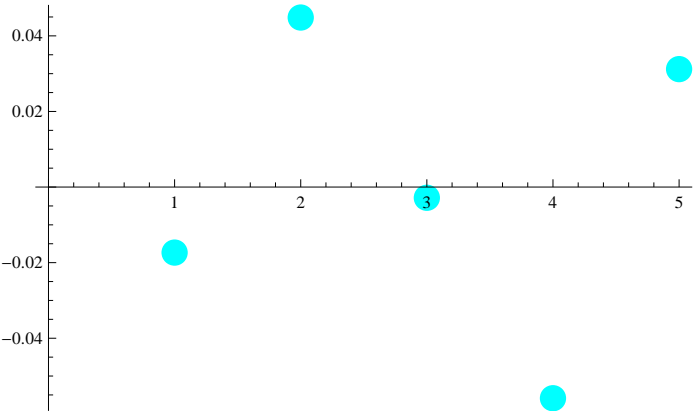
0.00213844, {-0.0173389, 0.0448496, -0.0028397, -0.0558717, 0.0312008},

	Estimate	Standard Error	t Statistic	P-Value
Ko	$2.99552 \times 10^{14}$	$6.87889 \times 10^{-17}$	$4.35466 \times 10^{30}$	$2.6706 \times 10^{-92}$
Cf	0.998674	0.0206532	48.3544	0.0000194757
$\{\{2.99552 \times 10^{14}, 2.99552 \times 10^{14}\}, \{0.932946, 1.0644\}\},$				

	Estimate	Standard Error	Confidence Interval
Ko	$2.99552 \times 10^{14}$	$6.87889 \times 10^{-17}$	$\{2.99552 \times 10^{14}, 2.99552 \times 10^{14}\}$
Cf	0.998674	0.0206532	$\{0.932946, 1.0644\}$
$\{0.999986, 0.999976\}$			

valordeR=0.998082

{-0.0173389, 0.0448496, -0.0028397, -0.0558717, 0.0312008}



⋮

```

(* Anexo 8 b Processo II ,
Haciendo una estimacion del parametro cf de Farjas*)

Clear[Ko];
 $\alpha_p = 0.632;$ 
Na = 6.02 * 10^23;
Kb = 1.381 * 10^-23;
R = 0.00831;
Ener = 206.121;
TgKelvin[x_] := 273.15 + x;
 $\beta_1 = 5;$ 
 $\beta_2 = 10;$ 
 $\beta_3 = 15;$ 
 $\beta_4 = 20;$ 
 $\beta_5 = 30;$ 
T1 = 311.2;
T2 = 320.4;
T3 = 326;
T4 = 330;
T5 = 335.7;
rapienfria = { $\beta_1$ ,  $\beta_2$ ,  $\beta_3$ ,  $\beta_4$ ,  $\beta_5$ };
listatepgradCent = {T1, T2, T3, T4, T5};
TemperaturaKelvin = {};
Do[ti = listatepgradCent[[i]];
  TemperaturaKelvin = AppendTo[TemperaturaKelvin, TgKelvin[ti]],
  {i, 1, Length[listatepgradCent]}]
TemperaturaKelvin
listay = {};
Do[Te = TemperaturaKelvin[[i]]; flujo = rapienfria[[i]];
  listay = AppendTo[listay, {Te^-1, Log[flujo / Te^2]}],
  {i, 1, Length[TemperaturaKelvin]}]
listay
resul = NonlinearModelFit[listay,
  
$$-\frac{\text{Ener}}{R} * y + \text{Log}\left[\frac{R * Ko * Cf}{\text{Ener}}\right], \{\{Ko, 9.88979 * 10^{17}\}, \{Cf, 0.9\}\}, \{y\}]$$

Listasali = {};
Listasali = resul[{"BestFitParameters", "ANOVATable", "EstimatedVariance",
  "FitResiduals", "ParameterTable", "ParameterConfidenceIntervals",
  "ParameterConfidenceIntervalTable", "RSquared", "AdjustedRSquared"}]
tablaAnov = resul["ANOVATable"];
valorSS = tablaAnov[[1, 1, 3, 3]];
valorST = tablaAnov[[1, 1, 5, 3]];

Print["valordeR=",  $\sqrt{1 - \frac{\text{valorSS}}{\text{valorST}}}$ ]

Listares = {};
Listares = Listasali[[4]]
ListPlot[Listares, PlotStyle -> {Hue[0.5], PointSize[0.04]}]

{584.35, 593.55, 599.15, 603.15, 608.85}

```

{ {0.0017113, -11.1316}, {0.00168478, -10.4697},  
{0.00166903, -10.083}, {0.00165796, -9.8086}, {0.00164244, -9.42195} }

FittedModel [ 31.3167- 24804.y ]

{ {Ko → 9.79972 × 10<sup>17</sup>, Cf → 1.00915} ,

	DF	SS	MS
Model	2	520.173	260.087
Error	3	0.0000121801	4.06004 × 10 <sup>-6</sup> , 4.06004 × 10 <sup>-6</sup> ,
Uncorrected Total	5	520.173	
Corrected Total	4	1.71133	

{ -0.00110045, 0.00287557, -0.0010261, -0.00120119, 0.000452166} ,

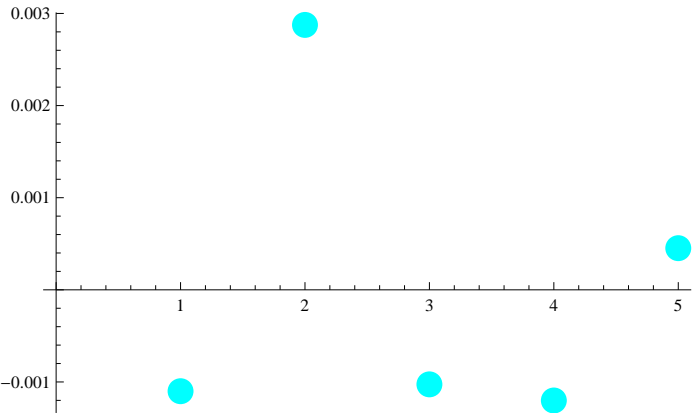
	Estimate	Standard Error	t Statistic	P-Value
Ko	9.79972 × 10 <sup>17</sup>	0.	∞	0. × 10 <sup>-308</sup> ,
Cf	1.00915	0.000909361	1109.74	1.61365 × 10 <sup>-9</sup>

{ {9.79972 × 10<sup>17</sup>, 9.79972 × 10<sup>17</sup>} , {1.00626, 1.01205} } ,

	Estimate	Standard Error	Confidence Interval
Ko	9.79972 × 10 <sup>17</sup>	0.	{9.79972 × 10 <sup>17</sup> , 9.79972 × 10 <sup>17</sup> } , 1. , 1. }
Cf	1.00915	0.000909361	{1.00626, 1.01205}

valordeR=0.999996

{ -0.00110045, 0.00287557, -0.0010261, -0.00120119, 0.000452166} }





## ■ (\* Anexo 9 Hallando a n Primer Proceso \*)

```

Ene = 116.969`;
R = 0.00831`; Ko = 2.85 * 10^14; Tp = 428.5; β = 20; Cf = 1;
Clear[n]
P[x_] := 
$$\frac{N[\text{ExpIntegralE}[2, x]]}{x}; Vx = \frac{Ene}{R Tp};$$

f1[n_] = -n (Ko Cf)^n 
$$\left(\frac{Ene}{R \beta}\right)^{n-1} P[Vx]^n \text{Exp}[-Vx] + \frac{(n-1) (\beta R) \text{Exp}[-Vx]}{Ene};$$

f2[x_] = -
$$\frac{\beta Ene P[Vx]}{R Tp^2};$$

f1b[n_] = n Log[Ko Cf P[Vx]] + (n-1) Log
$$\left[\frac{Ene}{R \beta}\right] + \text{Log}[n] - Vx;$$

f2b[n_] = Log
$$\left[\frac{(\beta Ene) P[Vx]}{R Tp^2} + \frac{(n-1) (\beta R) \text{Exp}[-Vx]}{Ene}\right];$$

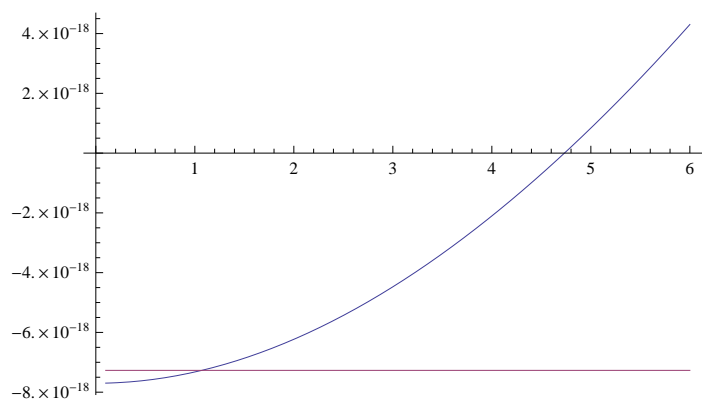
Plot[{f1[x], f2[x]}, {x, 0.1, 6}, PlotRange → All]
Plot[{f1b[x], f2b[x]}, {x, 0.1, 6}, PlotRange → All]
FindRoot[

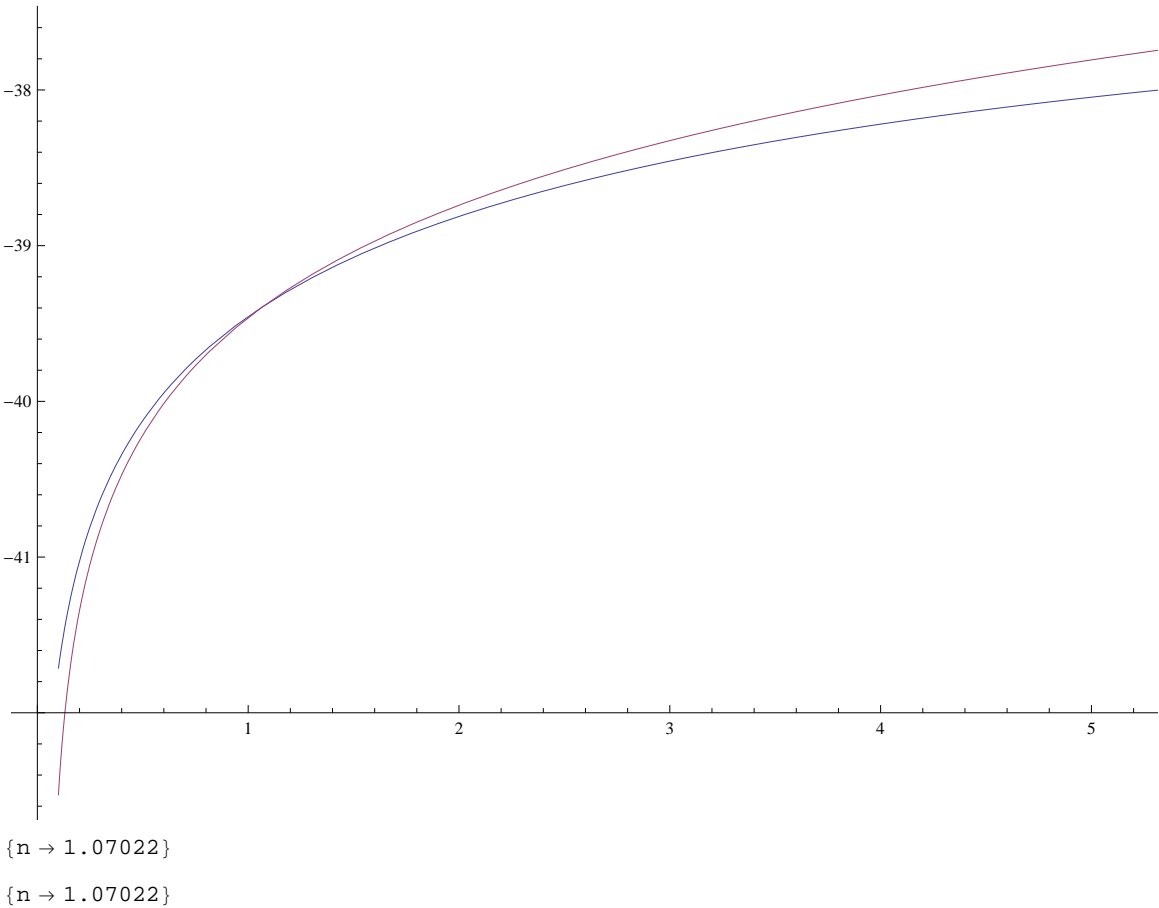
$$-n (Ko Cf)^n \left(\frac{Ene}{R \beta}\right)^{n-1} P[Vx]^n \text{Exp}[-Vx] + \frac{(n-1) (\beta R) \text{Exp}[-Vx]}{Ene} == -\frac{\beta Ene P[Vx]}{R Tp^2}, \{n, 1\}]$$

FindRoot[n Log[Ko Cf P[Vx]] + (n-1) Log
$$\left[\frac{Ene}{R \beta}\right] + \text{Log}[n] - Vx ==$$


$$\text{Log}\left[\frac{(\beta Ene) P[Vx]}{R Tp^2} + \frac{(n-1) (\beta R) \text{Exp}[-Vx]}{Ene}\right], \{n, 1\}]$$


```





■ (\* Anexo 9 b Programa para calcular os erros de n ,atualizado para  $\beta=5$  Processo II\*)

```
n = 0.66;
Ene = 206.211;
R = 0.00831;
Ko = 9.51 * 10 ^ 17;
Tp = 584.35;
β = 5;
Cf = 1.00;
DeltaT = 0.1;
δEne = 0.2;
```

```
N[ExpIntegralE[2, x]]
```

```
P[x_] :=  $\frac{\text{Ene}}{x}$ 
```

```
Vx =  $\frac{\text{Ene}}{R * \text{Tp}}$  ;
```

```
deltan =
```

$$\left( n * (Ko * Cf)^n * \left( \frac{\text{Ene}}{\beta * R} \right)^{n-1} * \text{Exp}[-Vx] * P[Vx]^n * \frac{\text{Ene}}{R * \text{Tp}^2} * \left( \frac{n * (P[Vx])^{-1} * \text{Exp}[-Vx]}{Vx^2} + 1 \right) + \frac{\beta * \text{Exp}[-Vx]}{\text{Tp}^2} * n + \frac{2 * \beta * \text{Ene} * P[Vx]}{R * \text{Tp}^3} \right) /$$

$$\left( \left( \frac{\text{Ene}}{\beta * R} \right)^{n-1} * P[Vx]^n * \text{Exp}[-Vx] * (Ko * Cf)^n * \left( 1 + \text{Log} \left[ \left( \frac{Ko * Cf * \text{Ene} * P[Vx]}{\beta * R} \right)^n \right] \right) + \frac{\beta * R * \text{Exp}[-Vx]}{\text{Ene}} \right) * \text{DeltaT} + \left( \left( (Ko * Cf)^n * \left( \frac{\text{Ene}}{\beta * R} \right)^{n-1} * \text{Exp}[-Vx] * P[Vx]^n * \left( \frac{n * (n-1)}{\text{Ene}} + \frac{n^2 * (P[Vx])^{-1} * \text{Exp}[-Vx] * R * \text{Tp}}{\text{Ene}^2} + \frac{n}{R * \text{Tp}} \right) + \frac{(n-1) * \beta * R * \text{Exp}[-Vx]}{\text{Ene}} * \left( \frac{1}{\text{Ene}} + \frac{1}{R * \text{Tp}} \right) + \frac{\beta * P[Vx]}{R * \text{Tp}^2} + \frac{\beta * \text{Exp}[-Vx]}{\text{Ene} * \text{Tp}} \right) * \delta \text{Ene} \right) /$$

$$\left( \left( \frac{\text{Ene}}{\beta * R} \right)^{n-1} * P[Vx]^n * \text{Exp}[-Vx] * (Ko * Cf)^n * \left( 1 + \text{Log} \left[ \left( \frac{Ko * Cf * \text{Ene} * P[Vx]}{\beta * R} \right)^n \right] \right) + \frac{\beta * R * \text{Exp}[-Vx]}{\text{Ene}} \right)$$

```
n + deltan
```

```
n - deltan
```

```
0.0444671
```

```
0.704467
```

```
0.615533
```

## ■ (\* Anexo 10 Calculo Dos Erros en KO \*)

```

In[1]:= Ene = 116.969;
Cf = 1.0;
β = 10;
R = 0.00831;
Ti = 418.95;
deltaCf = 0.01;
deltaEne = 0.2;
deltaP = 5 * 10^-19;
ListaLeffTTeor = {};
listLeffTexpS = {};

N[ExpIntegrale[2,  $\frac{Ene}{R * T}$ ]]

Pt[T_] =  $\frac{Ene}{R * T}$ ;

Pt[Ti];

valorKO =  $\frac{R * \beta}{Cf * Ene * Pt[Ti]}$ 

deltaKo2 = R * β *  $\left( \frac{deltaCf}{(Cf^2) * Ene * Pt[Ti]} + \frac{deltaP}{(Pt[Ti]^2) * Ene * Cf} + \frac{deltaEne}{Cf * Pt[Ti] * Ene^2} \right)$ 

Print["Ko+δKo2 = ", valorKO + deltaKo2]
Print["Ko-δKo2 = ", valorKO - deltaKo2]

```

Out[13]=  $3.31005 \times 10^{14}$

Out[14]=  $8.09857 \times 10^{13}$

Ko+δKo2 =  $4.11991 \times 10^{14}$

Ko-δKo2 =  $2.50019 \times 10^{14}$

■ (\* Anexo 10 b Calculo Dos Erros en KO \*)

$$\text{deltaKo} = \frac{R * \beta}{\text{Pt}[\text{Ti}]} * \left( \frac{\text{deltaCf}}{(\text{Cf}^2) * \text{Ene}} + \frac{\text{deltaEne}}{\text{Cf} * \text{Ene}^2} \right);$$

Print["Ko+δKo = ", valorKO + deltaKo]

Print["Ko-δKo = ", valorKO - deltaKo]

Ene = 206.211;

Cf = 1.0;

β = 30;

R = 0.00831;

Ti = 608.85;

deltaCf = 0.01;

deltaEne = 0.2;

deltaP = 5 \* 10<sup>-22</sup>;

ListaLefFTTeor = {};

listLefFTexpS = {};

$$N\left[\text{ExpIntegraleE}\left[2, \frac{\text{Ene}}{R * T}\right]\right]$$

$$\text{Pt}[\text{T}_-] = \frac{\text{Ene}}{R * T};$$

Pt[Ti];

$$\text{valorKO} = \frac{R * \beta}{\text{Cf} * \text{Ene} * \text{Pt}[\text{Ti}]}$$

$$\text{deltaKo2} = R * \beta * \left( \frac{\text{deltaCf}}{(\text{Cf}^2) * \text{Ene} * \text{Pt}[\text{Ti}]} + \frac{\text{deltaP}}{(\text{Pt}[\text{Ti}]^2) * \text{Ene} * \text{Cf}} + \frac{\text{deltaEne}}{\text{Cf} * \text{Pt}[\text{Ti}] * \text{Ene}^2} \right)$$

Print["Ko+δKo2 = ", valorKO + deltaKo2]

Print["Ko-δKo2 = ", valorKO - deltaKo2]

$$1.05592 \times 10^{18}$$

$$4.72712 \times 10^{17}$$

$$\text{Ko} + \delta\text{Ko2} = 1.52863 \times 10^{18}$$

$$\text{Ko} - \delta\text{Ko2} = 5.83209 \times 10^{17}$$

## Anexo 11

(\*O programa usado para calcular no. Neste programa  
 $\alpha$  (T) é a função de interpolação obtida da data experimental ,  
a mesma é diferente para cada taxa de aquecimento  $\beta$  \*)

In[28]:= **Ene = 116.969;**

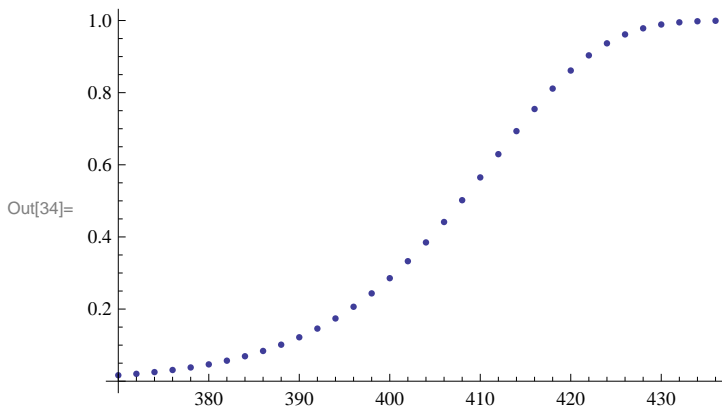
**Ko = 2.99184 \* 10<sup>14</sup>;**

**Cf = 1.0;**

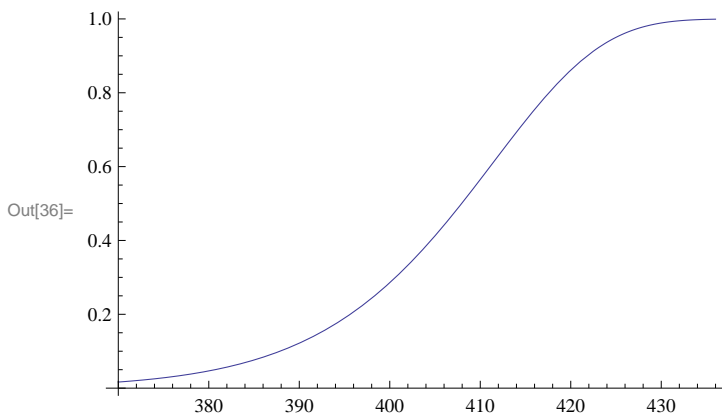
**$\beta$  = 5;**

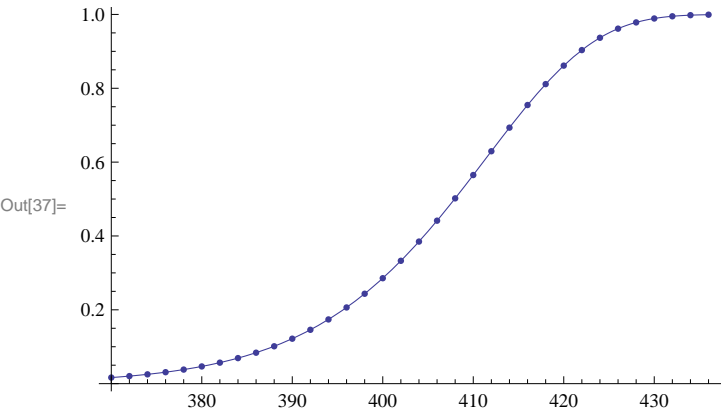
**R = 0.00831;**

In[33]:= **Listaexpft = {{370, 0.01651760622200371`}, {372, 0.02043987958258464`},**  
**{374, 0.025225522215615692`}, {376, 0.031046125340003616`},**  
**{378, 0.03810190930459556`}, {380, 0.046624631722749665`},**  
**{382, 0.056880022112787865`}, {384, 0.06916931528982617`},**  
**{386, 0.08382929198766287`}, {388, 0.10123003753123139`},**  
**{390, 0.12176940521334956`}, {392, 0.14586294026377278`}, {394, 0.1739278227326989`},**  
**{396, 0.20635929206972237`}, {398, 0.2434981299979856`}, {400, 0.2855882535426786`},**  
**{402, 0.33272449952435434`}, {404, 0.3847924728116796`}, {406, 0.44140504078160203`},**  
**{408, 0.5018436741797023`}, {410, 0.5650169974865198`}, {412, 0.6294526878466693`},**  
**{414, 0.6933405814716751`}, {416, 0.754642170512299`}, {418, 0.8112721570151054`},**  
**{420, 0.8613401219962658`}, {422, 0.9034165591934372`}, {424, 0.9367643587650922`},**  
**{426, 0.9614660315452251`}, {428, 0.9783911188275345`}, {430, 0.9889923252021308`},**  
**{432, 0.9949813145170145`}, {434, 0.9979864940641315`}, {436, 0.9993027587390773`}};**  
**gflistexp = ListPlot[Listaexpft]**  
 **$\alpha$  = Interpolation[Listaexpft]**  
**gfinterft = Plot[ $\alpha$ [x], {x, 370., 436.}]**  
**Show[gfinterft, gflistexp]**



Out[35]= **InterpolatingFunction[{{370., 436.}}, <>]**





```
In[38]:= fnte[T_] = 1 -  $\alpha$ [T];
derivadafnt[T_] =  $\partial_T$ (fnte[T]);

LeftFTTTeor[T_, no_] =  $\left(-\text{derivadafnt}[T] * \frac{\beta}{K_0} * \text{Exp}\left[\frac{E_{ne}}{R * T}\right]\right)^{\left(\frac{1}{no}\right)}$ 

Out[40]=  $1.67121 \times 10^{-14} \frac{1}{no} \left(e^{14075.7/T} \text{InterpolatingFunction}[\{\{370., 436.\}\}, <>][T]\right)^{\frac{1}{no}}$ 

In[41]:= Listaexp = {};
Do[Listaexp = AppendTo[Listaexp, {T, fnte[T]}], {T, 370, 436, 2}]

In[43]:= ajuste = NonlinearModelFit[Listaexp, LeftFTTTeor[T, no], {{no, 0.5}}, {T}]

Listasali = {};
Listasali = ajuste[{"BestFitParameters", "ANOVATable", "EstimatedVariance",
"FitResiduals", "ParameterTable", "ParameterConfidenceIntervals",
"ParameterConfidenceIntervalTable", "RSquared", "AdjustedRSquared"}]
```

Out[43]= FittedModel [

$1.68459 \times 10^{-14} \left(e^{14075.7/T} \text{InterpolatingFunction}[\{\{370., 436.\}\}, <>][T]\right)^{0.999749}$

]

Out[45]= { {no → 1.00025},

	DF	SS	MS
Model	1	14.8838	14.8838
Error	33	$4.75437 \times 10^{-6}$	$1.44072 \times 10^{-7}$
Uncorrected Total	34	14.8838	
Corrected Total	33	4.71203	

$1.44072 \times 10^{-7}, \{-0.0018203, 0.000500081, -0.000428471, -0.000391476, -0.00035212,$   
 $-0.000310149, -0.000265375, -0.000217722, -0.000167283, -0.000114383, -0.0000596717,$   
 $-4.2072 \times 10^{-6}, 0.0000504468, 0.000102146, 0.000148132, 0.000185079, 0.000209271,$   
 $0.00021696, 0.000204945, 0.000171386, 0.000116764, 0.0000448067, -0.0000369459,$   
 $-0.000117398, -0.000183372, -0.000222449, -0.000226709, -0.000196155, -0.000140031,$   
 $-0.0000746016, -0.0000174813, 0.0000191555, 0.0000327884, -0.0000958284\},$

	Estimate	Standard Error	t Statistic	P-Value
no	1.00025	0.000319933	3126.44	$7.21434 \times 10^{-92}$
	Estimate	Standard Error	Confidence Interval	
no	1.00025	0.000319933	{0.9996, 1.0009}	

, {{0.9996, 1.0009}}, {1., 1.}

## ■ Anexo 12(\*calculodeEFriedmanIprocesom7.nb\*)

```
In[1]:= R = 0.00831;
TgKelvin[x_] := 273.15 + x;
φ1 = 5;
φ2 = 10;
φ3 = 15;
φ4 = 20;
φ5 = 30;
T1 = 138.5;
T2 = 145.8;
T3 = 151.2;
T4 = 155.35;
T5 = 159.3
vdyi1 = 4.431 * 10^-6;
vdyi2 = 4.95 * 10^-6;
vdyi3 = 4.463 * 10^-6;
vdyi4 = 4.494 * 10^-6;
vdyi5 = 4.95 * 10^-6;
rapienfria = {φ1, φ2, φ3, φ4, φ5};
valoresdeyie = {vdyi1, vdyi2, vdyi3, vdyi4, vdyi5};
listatepgradCent = {T1, T2, T3, T4, T5};
TemperaturaKelvin = {};
Do[ti = listatepgradCent[[i]];
  TemperaturaKelvin = AppendTo[TemperaturaKelvin, TgKelvin[ti]],
  {i, 1, Length[listatepgradCent]}]
TemperaturaKelvin
listay = {};
Do[Te = TemperaturaKelvin[[i]]; valordey = valoresdeyie[[i]];
  listay = AppendTo[listay, {Te^-1, -Log[rapienfria[[i]] * valordey]}],
  {i, 1, Length[TemperaturaKelvin]}]
listay
resul = NonlinearModelFit[listay, a + b * x, {a, b}, {x}]
l = Normal[resul]
NonlinearModelFit[listay, a + b * x, {a, b}, {x}]
Print["Pendiente =", " ", pend = l[[2, 1]]]
Print["intercepto =", " ", Interc = l[[1]]]
Print["Energiaactiva=", " ", Energiaactiva1 = pend * R]
```

Out[12]= 159.3

Out[23]= {411.65, 418.95, 424.35, 428.5, 432.45}

Out[26]= {{0.00242925, 10.7174}, {0.00238692, 9.91354},  
{0.00235655, 9.61164}, {0.00233372, 9.31704}, {0.00231241, 8.81493}}

Out[27]= FittedModel  $\left[ -26.602 + 15\,347.1 x \right]$

Out[28]=  $-26.602 + 15\,347.1 x$

Out[29]= FittedModel  $\left[ -26.602 + 15\,347.1 x \right]$

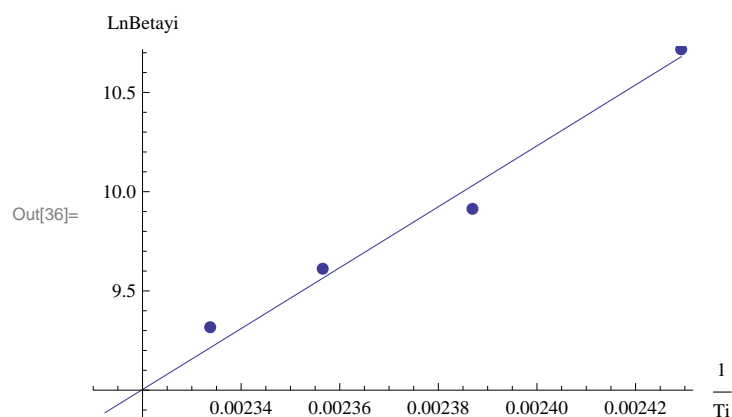
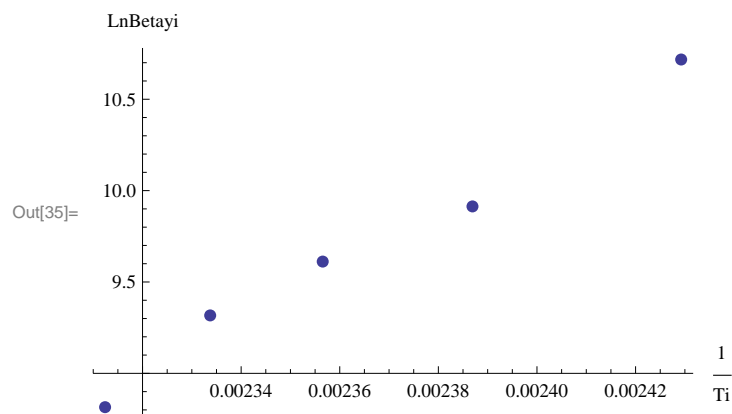
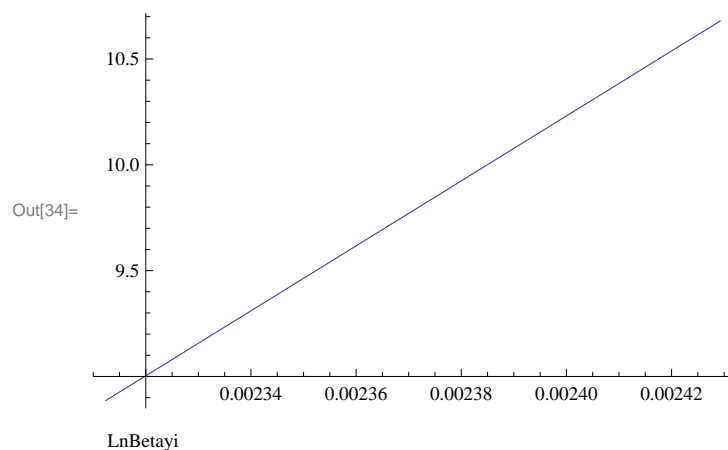


Pendiente = 15 347.1

intercepto = -26.602

Energiaactiva= 127.534

```
In[33]:= ajuste[x_] := Interc + pend * x
Gf1 = Plot[ajuste[x], {x, 0.0024292481476982874, 0.0023124060585038735`}]
Gf2 = ListPlot[listay, PlotStyle -> PointSize [0.02], AxesLabel -> {Ti^-1, LnBetayi}]
Show[Gf1, Gf2, AxesLabel -> {Ti^-1, LnBetayi}]
NonlinearModelFit[listay, a + b x, {a, b}, {x}] [
{"ANOVATable", "ParameterPValues", "AdjustedRSquared", "RSquared",
"ParameterTStatistics", "ParameterConfidenceIntervalTable "}]
```



	DF	SS	MS
Model	2	470.002	235.001
Error	3	0.0330961	0.011032
Uncorrected Total	5	470.036	
Corrected Total	4	2.01548	

	Estimate	Standard Error	Confidence Interval
0.99993, {-9.82844, 13.405}, a	-26.602	2.70663	{-35.2157, -17.9883}
b	15 347.1	1144.88	{11 703.5, 18 990.6}

```

In[38]:= lm = LinearModelFit[listay, x, x]
listasalida = {};
listasalida = lm[{"AdjustedRSquared", "RSquared", "FitResiduals"}]
 $\sqrt{\text{listasalida}[[2]]}$ 
Listares = {};
Listares = listasalida[[3]]
ListPlot[Listares, PlotStyle -> {Hue[0.5], PointSize[0.04]}]
listasalida2 = {};
listasalida2 = lm[{"FitResiduals",
  "SinglePredictionConfidenceIntervalTable", "ParameterConfidenceRegion"}];
errors = listasalida2[[1]]
tablasimpledeintecnf = listasalida2[[2]]
Observed = {}
Do[Observed = AppendTo[Observed, tablasimpledeintecnf[[1, 1, i, 1]]], {i, 2, 6}]
Observed
predicted = {};
Do[predicted = AppendTo[predicted, tablasimpledeintecnf[[1, 1, i, 2]]], {i, 2, 6}]
predicted
ListPlot[Transpose[{predicted, errors}], PlotStyle -> PointSize[0.02]]
gplaj = Plot[lm[x], {x, 0.0024292481476982874, 0.0023124060585038735}];
ci = {}
Do[ci = AppendTo[ci, tablasimpledeintecnf[[1, 1, i, 4]]], {i, 2, 6}]
ci
(xval = First /@ listay; predicted = Transpose[{xval, predicted}];
  lowerCI = Transpose[{xval, First /@ ci}]; upperCI = Transpose[{xval, Last /@ ci}]);
gfrs = ListPlot[{listay, predicted, lowerCI, upperCI}, Joined -> {False, True, True, True},
  PlotStyle -> {Automatic, Automatic, Dashing[{0.05}, 0.05]}, Dashing[{0.05}, 0.05]}]
Show[gfrs, gplaj, AxesLabel -> {Ti^-1, LnBetayi}]

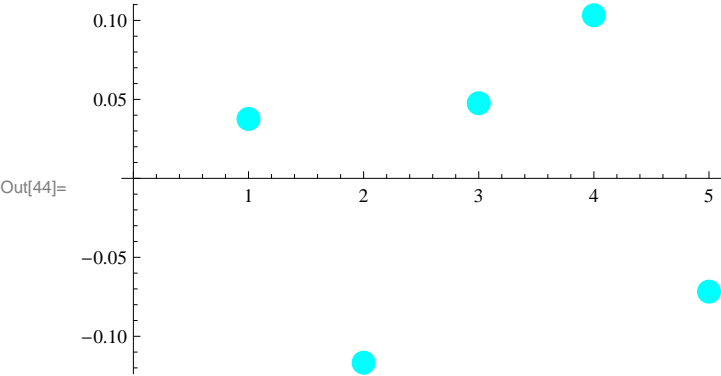
```

Out[38]= FittedModel  $\left[ -26.602 + 15\,347.1 x \right]$

Out[40]= {0.978105, 0.983579, {0.0376074, -0.116685, 0.0475738, 0.103236, -0.0717323}}

Out[41]= 0.991756

Out[43]= {0.0376074, -0.116685, 0.0475738, 0.103236, -0.0717323}



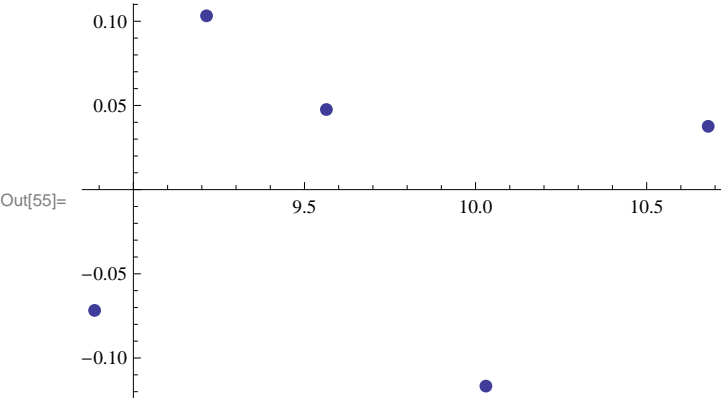
Out[47]= {0.0376074, -0.116685, 0.0475738, 0.103236, -0.0717323}

	Observed	Predicted	Standard Error	Confidence Interval
Out[48]=	10.7174	10.6798	0.137326	{10.2428, 11.1169}
	9.91354	10.0302	0.118072	{9.65447, 10.406}
	9.61164	9.56407	0.115355	{9.19695, 9.93118}
	9.31704	9.2138	0.12009	{8.83162, 9.59598}
	8.81493	8.88666	0.129214	{8.47544, 9.29787}

Out[49]= {}

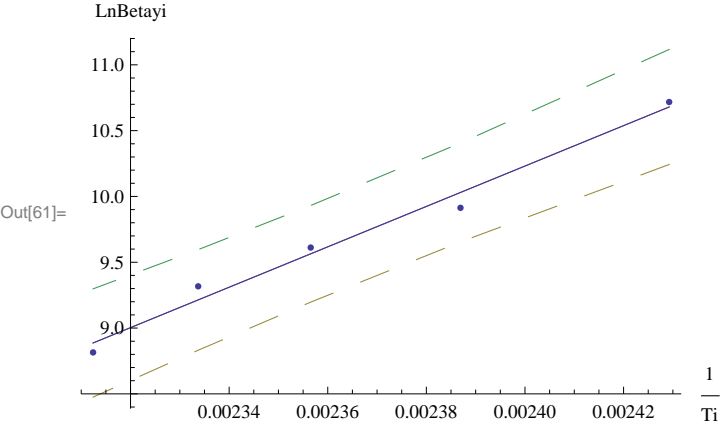
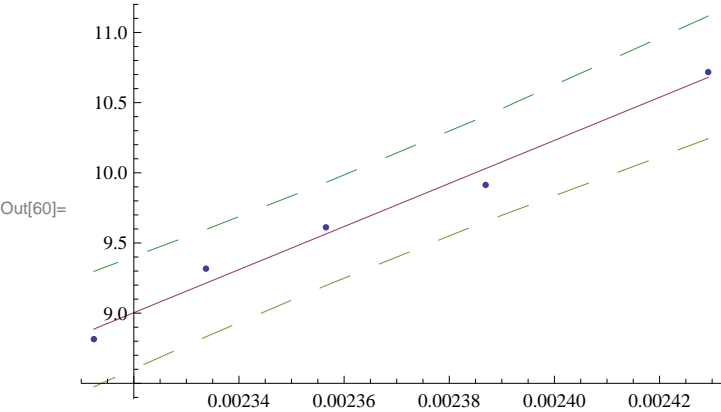
Out[51]= {10.7174, 9.91354, 9.61164, 9.31704, 8.81493}

Out[54]= {10.6798, 10.0302, 9.56407, 9.2138, 8.88666}



Out[57]= {}

Out[59]= {{10.2428, 11.1169}, {9.65447, 10.406},  
{9.19695, 9.93118}, {8.83162, 9.59598}, {8.47544, 9.29787}}



## ■ Processo II

```

R = 0.00831;
TgKelvin[x_] := 273.15 + x;
φ1 = 5;
φ2 = 10;
φ3 = 15;
φ4 = 20;
φ5 = 30;
T1 = 311.2;
T2 = 320.4;
T3 = 326;
T4 = 330;
T5 = 335.5
vdyi1 = 5.083 * 10^-6;
vdyi2 = 5.606 * 10^-6;
vdyi3 = 5.349 * 10^-6;
vdyi4 = 4.822 * 10^-6;
vdyi5 = 4.736 * 10^-6;
rapienfria = {φ1, φ2, φ3, φ4, φ5};
valoresdeyie = {vdyi1, vdyi2, vdyi3, vdyi4, vdyi5};
listatepgradCent = {T1, T2, T3, T4, T5};
TemperaturaKelvin = {};
Do[ti = listatepgradCent[[i]];
  TemperaturaKelvin = AppendTo[TemperaturaKelvin, TgKelvin[ti]],
  {i, 1, Length[listatepgradCent]}]
TemperaturaKelvin
listay = {};
Do[Te = TemperaturaKelvin[[i]]; valordey = valoresdeyie[[i]];
  listay = AppendTo[listay, {Te^-1, -Log[rapienfria[[i]] * valordey]}],
  {i, 1, Length[TemperaturaKelvin]}]
listay
resul = NonlinearModelFit[listay, a + b * x, {a, b}, {x}]
l = Normal[resul]
NonlinearModelFit[listay, a + b * x, {a, b}, {x}]
Print["Pendiente =", " ", pend = l[[2, 1]]]
Print["intercepto =", " ", Interc = l[[1]]]
Print["Energiaactiva=", " ", Energiaactiva1 = pend * R]

```

335.5

{584.35, 593.55, 599.15, 603.15, 608.65}

{0.0017113, 10.5802}, {0.00168478, 9.78909},  
 {0.00166903, 9.43055}, {0.00165796, 9.24659}, {0.00164298, 8.85912}}

FittedModel[  $-31.8238 + 24\,745.8 x$  ]

$-31.8238 + 24\,745.8 x$

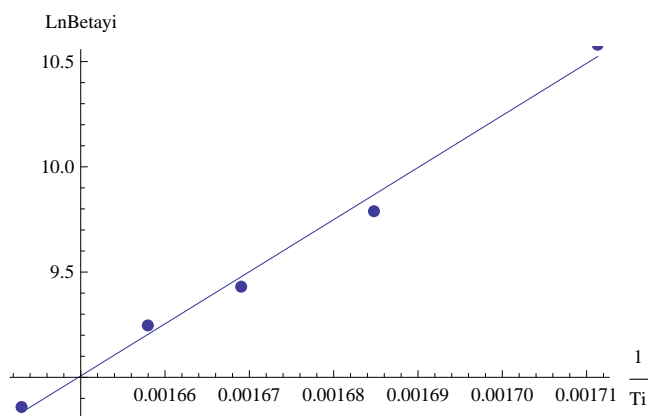
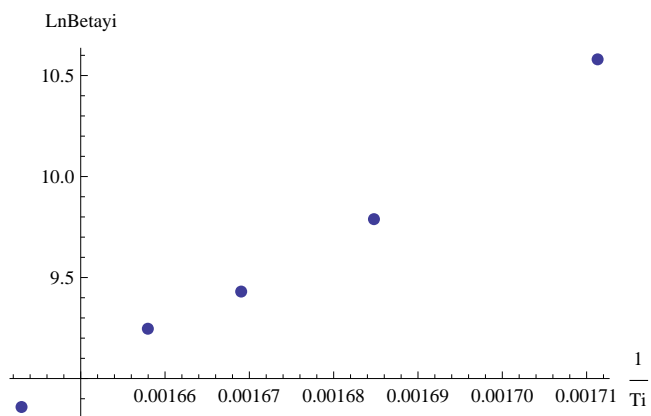
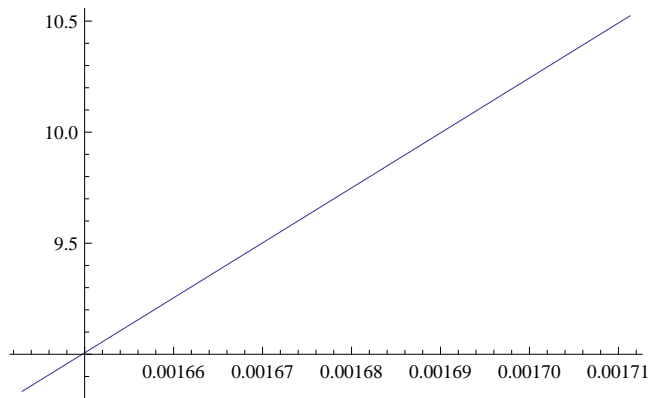
FittedModel[  $-31.8238 + 24\,745.8 x$  ]

Pendiente = 24745.8

intercepto = -31.8238

Energiaactiva= 205.638

```
ajuste[x_] := Interc + pend * x
Gf1 = Plot[ajuste[x], {x, 0.0017113031573543255, 0.0016429803663846217}]
Gf2 = ListPlot[listay, PlotStyle -> PointSize [0.02], AxesLabel -> {Ti^-1, LnBetayi}]
Show[Gf1, Gf2, AxesLabel -> {Ti^-1, LnBetayi}]
NonlinearModelFit[listay, a + b x, {a, b}, {x}] [
  {"ANOVATable", "ParameterPValues", "AdjustedRSquared", "RSquared",
   "ParameterTStatistics", "ParameterConfidenceIntervalTable"}]
```



	DF	SS	MS
Model	2	460.671	230.335
{ Error	3	0.0140447	0.00468157
Uncorrected Total	5	460.685	
Corrected Total	4	1.69722	

{0.000700773, 0.000320283}, 0.999949, 0.99997, {-14.5722, 18.9613},

	Estimate	Standard Error	Confidence Interval
a	-31.8238	2.18387	{-38.7739, -24.8738}
b	24 745.8	1305.07	{20 592.5, 28 899.1}

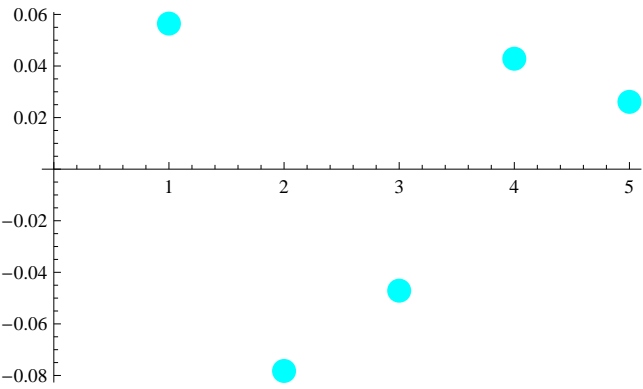
```
lm = LinearModelFit[listay, x, x]
listasalida = {};
listasalida = lm[{"AdjustedRSquared", "RSquared", "FitResiduals"}]
 $\sqrt{\text{listasalida}[[2]]}$ 
Listares = {};
Listares = listasalida[[3]]
ListPlot[Listares, PlotStyle -> {Hue[0.5], PointSize[0.04]}]
listasalida2 = {};
listasalida2 = lm[{"FitResiduals",
  "SinglePredictionConfidenceIntervalTable", "ParameterConfidenceRegion"}];
errors = listasalida2[[1]]
tablasimpledeintecnf = listasalida2[[2]]
Observed = {}
Do[Observed = AppendTo[Observed, tablasimpledeintecnf[[1, 1, i, 1]]], {i, 2, 6}]
Observed
predicted = {};
Do[predicted = AppendTo[predicted, tablasimpledeintecnf[[1, 1, i, 2]]], {i, 2, 6}]
predicted
ListPlot[Transpose[{predicted, errors}], PlotStyle -> PointSize[0.02]]
gplaj = Plot[lm[x], {x, 0.0017113031573543255, 0.0016429803663846217}];
ci = {}
Do[ci = AppendTo[ci, tablasimpledeintecnf[[1, 1, i, 4]]], {i, 2, 6}]
ci
(xval = First /@ listay; predicted = Transpose[{xval, predicted}];
  lowerCI = Transpose[{xval, First /@ ci}]; upperCI = Transpose[{xval, Last /@ ci}]);
gfrs = ListPlot[{listay, predicted, lowerCI, upperCI}, Joined -> {False, True, True, True},
  PlotStyle -> {Automatic, Automatic, Dashing[{0.05}, 0.05]}, Dashing[{0.05}, 0.05]}]
Show[gfrs, gplaj, AxesLabel -> {Ti^-1, LnBetayi}]
```

FittedModel [  $-31.8238 + 24\,745.8 x$  ]

{0.988966, 0.991725, {0.0564464, -0.0782511, -0.0471187, 0.0428255, 0.0260978}}

0.995854

{0.0564464, -0.0782511, -0.0471187, 0.0428255, 0.0260978}



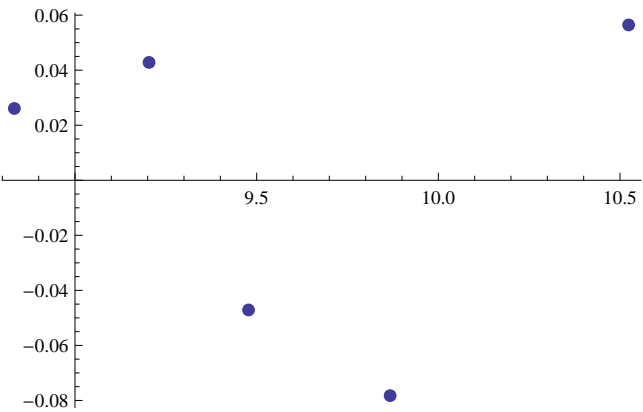
{0.0564464, -0.0782511, -0.0471187, 0.0428255, 0.0260978}

Observed	Predicted	Standard Error	Confidence Interval
10.5802	10.5237	0.0899403	{10.2375, 10.81}
9.78909	9.86734	0.0764576	{9.62402, 10.1107}
9.43055	9.47767	0.0751508	{9.23851, 9.71683}
9.24659	9.20376	0.0775495	{8.95697, 9.45056}
8.85912	8.83302	0.084702	{8.56346, 9.10258}

{}

{10.5802, 9.78909, 9.43055, 9.24659, 8.85912}

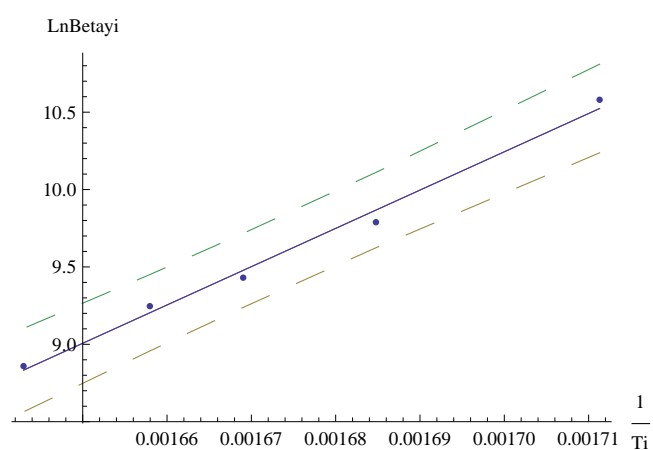
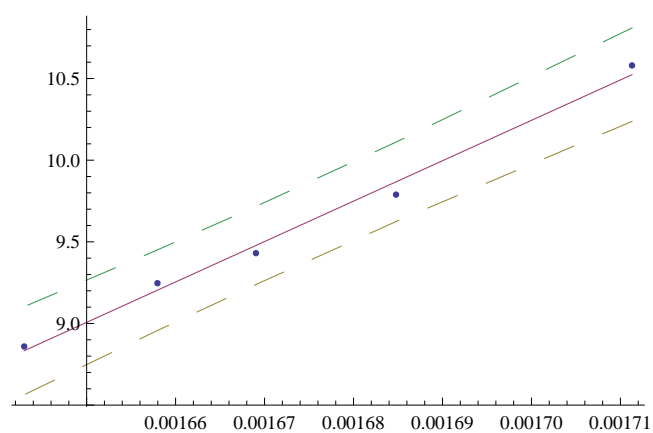
{10.5237, 9.86734, 9.47767, 9.20376, 8.83302}



{}

{{10.2375, 10.81}, {9.62402, 10.1107},  
{9.23851, 9.71683}, {8.95697, 9.45056}, {8.56346, 9.10258}}





r

## ■ Anexo 13 Hallando a n(T) y K(T)

```

dTdq1 = 0.382727;
dTdq2 = 0.81325;
dTdq3 = 1.29671;
dTdq4 = 1.85862;
dTdq5 = 2.52029;
dTdq6 = 3.33453;
dTdq7 = 4.38734;
ξ1 = 0.1;
ξ2 = 0.2;
ξ3 = 0.3;
ξ4 = 0.4;
ξ5 = 0.5;
ξ6 = 0.6;
ξ7 = 0.7;

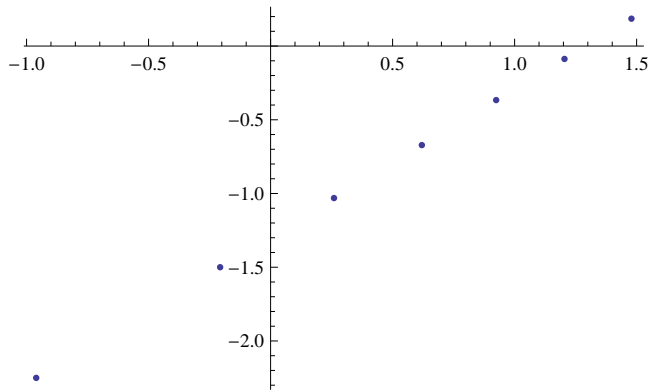
Listadedtdq = {dTdq1, dTdq2, dTdq3, dTdq4, dTdq5, dTdq6, dTdq7}
ListadeFraT = {ξ1, ξ2, ξ3, ξ4, ξ5, ξ6, ξ7}
Listatrabajo = {};
Do[Listatrabajo =
    AppendTo[Listatrabajo, {Log[Listadedtdq[[i]]], Log[Log[ $\frac{1}{1 - \text{ListadeFraT}[[i]]}$ ]]}],
    {i, 1, Length[ListadeFraT]}]
Listatrabajo
ListPlot[Listatrabajo]
resul =
    NonlinearModelFit[Listatrabajo, interc + pendt * x, {{interc, -0.1}, {pendt, 3}}, {x}]
Listasali = {};
Listasali = resul[{"BestFitParameters", "ANOVATable", "EstimatedVariance",
    "FitResiduals", "ParameterTable", "ParameterConfidenceIntervals",
    "ParameterConfidenceIntervalTable", "RSquared", "AdjustedRSquared"}]
tablaAnov = resul["ANOVATable"];
valorSS = tablaAnov[[1, 1, 3, 3]];
valorST = tablaAnov[[1, 1, 5, 3]];

Print["valordeR=",  $\sqrt{1 - \frac{\text{valorSS}}{\text{valorST}}}$ ]

Listares = {};
Listares = Listasali[[4]]
ListPlot[Listares, PlotStyle → {Hue[0.5], PointSize[0.02]}]
{0.382727, 0.81325, 1.29671, 1.85862, 2.52029, 3.33453, 4.38734}
{0.1, 0.2, 0.3, 0.4, 0.5, 0.6, 0.7}

{{-0.960433, -2.25037}, {-0.206717, -1.49994},
{0.25983, -1.03093}, {0.619834, -0.671727},
{0.924374, -0.366513}, {1.20433, -0.0874216}, {1.47872, 0.185627}}

```



FittedModel [  $-1.2913 + 0.999365 x$  ]

{ {interc → -1.2913, pendt → 0.999365},

	DF	SS	MS
Model	2	9.00443	4.50222
Error	5	$7.13862 \times 10^{-6}$	$1.42772 \times 10^{-6}$ , $1.42772 \times 10^{-6}$ ,
Uncorrected Total	7	9.00444	
Corrected Total	6	4.3283	

{0.000756846 , -0.00205363 , 0.000705308 , 0.000133459 , 0.0010013 , 0.000312728 ,

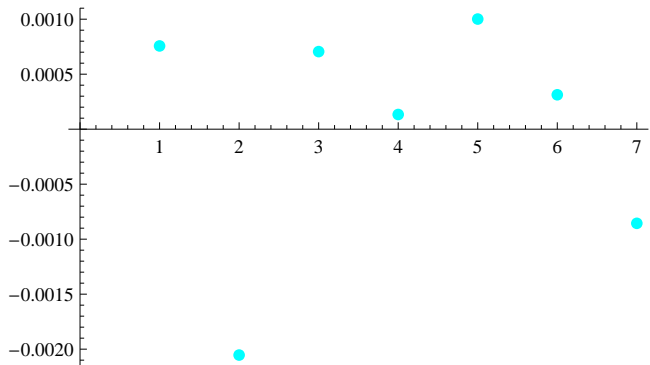
		Estimate	Standard Error	t Statistic	P-Value
-0.000856011},	interc	-1.2913	0.000527318	-2448.81	$2.1554 \times 10^{-16}$ ,
	pendt	0.999365	0.000573968	1741.15	$1.1861 \times 10^{-15}$

{ {-1.29266 , -1.28995} , {0.997889 , 1.00084} } ,

	Estimate	Standard Error	Confidence Interval
interc	-1.2913	0.000527318	{-1.29266 , -1.28995} , 0.999999 , 0.999999 }
pendt	0.999365	0.000573968	{0.997889 , 1.00084}

valordeR=0.999999

{0.000756846 , -0.00205363 , 0.000705308 ,  
0.000133459 , 0.0010013 , 0.000312728 , -0.000856011 }



**varK**  
**valon**  
  
varK  
  
valon

**Anexo 14 a** Dilatação Relativa, Processo I $\beta = 5$ 

T	$l(T)$
387.871	0.0000359
395.631	0.0000718
400.636	0.0001077
404.554	0.0001436
407.942	0.0001795
411.089	0.0002154
414.214	0.0002513
417.787	0.0002872
421.823	0.0003231
430	0.000359

valor de  $L1 = (\delta l/l)_{finl} = 0.000359$  $\beta = 10$ 

T	$l(T)$
395.028	0.0000364
403.72	0.0000728
408.264	0.0001092
412.329	0.0001456
415.846	0.000182
419.113	0.0002184
422.358	0.0002548
425.861	0.0002912
430.263	0.0003276
439	0.000364

valor de  $L1 = (\delta l/l)_{finl} = 0.000364$  $\beta = 15$ 

T	$l(T)$
399.335	0.0000394
407.553	0.0000788
412.858	0.0001182
417.013	0.0001576
420.609	0.000197
423.949	0.0002364
427.268	0.0002758
430.851	0.0003152
435.355	0.0003546
444.4	0.000394

valor de  $L1 = (\delta l/l)_{finl} = 0.000394$

$$\beta = 20$$

T	$l(T)$
402.447	0.0000412
410.791	0.0000824
416.18	0.0001236
420.4	0.0001648
424.053	0.000206
427.447	0.0002472
430.819	0.0002884
434.461	0.0003296
439.039	0.0003708
447	0.000412

valor de  $L1 = (\delta l/l)_{finl} = 0.000412$

$$\beta = 30$$

T	$l(T)$
406.914	0.0000418
415.442	0.0000836
420.95	0.0001254
425.266	0.0001672
429.001	0.000209
432.473	0.0002508
435.924	0.0002926
439.65	0.0003344
444.335	0.0003762
454	0.000418

valor de  $L1 = (\delta l/l)_{finl} = 0.000418$

## Anexo 14 b Dilatação Relativa, Processo II

$$\beta = 5$$

T	$l(T)$
543	0.0000515
555.74	0.000103
564.355	0.0001545
571.13	0.000206
577.15	0.0002575
582.75	0.000309
587.97	0.0003605
593.89	0.000412
601.36	0.0004635
611	0.000515

$$\text{valor de } L1 = (\delta l/l)_{finl} = 0.000515$$

$$\beta = 10$$

T	$l(T)$
550.49	0.0000529
564.14	0.0001058
573.015	0.0001587
579.995	0.0002116
586.06	0.0002645
591.72	0.0003174
597.36	0.0003703
603.465	0.0004232
611.175	0.0004761
623	0.000529

$$\text{valor de } L1 = (\delta l/l)_{finl} = 0.000529$$

$$\beta = 15$$

T	$l(T)^2$
555.28	0.000054
569.17	0.000108
578.2	0.000162
585.31	0.000216
591.485	0.00027
597.245	0.000324
602.98	0.000378
609.21	0.000432
617.06	0.000486
627	0.00054

$$\text{valor de } L1 = (\delta l/l)_{finl} = 0.00054$$

$\beta = 20$ 

T	$l(T)$
558.73	0.0000588
572.79	0.0001176
581.935	0.0001764
589.135	0.0002352
595.39	0.000294
601.225	0.0003528
607.04	0.0004116
613.345	0.0004704
621.305	0.0005292
631.4	0.000588

valor de  $L1 = (\delta l / l)_{finl} = 0.000588$

 $\beta = 30$ 

T	$l(T)$
563.67	0.0000601
577.975	0.0001202
587.28	0.0001803
594.61	0.0002404
600.98	0.0003005
606.925	0.0003606
612.84	0.0004207
619.275	0.0004808
627.385	0.0005409
639	0.000601

valor de  $L1 = (\delta l / l)_{finl} = 0.000601$



- Anexo 15 Version CHT a TTT(\* e necesario ter processadas as listas das fracT Vs B antes de començar,ListaFTVST[q] \*)

- Armando (beta T)

```

ListaBT[0.1] = {};
ListaBT[0.2] = {};
ListaBT[0.3] = {};
ListaBT[0.4] = {};
ListaBT[0.5] = {};
ListaBT[0.6] = {};
ListaBT[0.7] = {};
ListaBT[0.8] = {};
ListaBT[0.9] = {};
ListaBTp[0.1] = {};
ListaBTp[0.2] = {};
ListaBTp[0.3] = {};
ListaBTp[0.4] = {};
ListaBTp[0.5] = {};
ListaBTp[0.6] = {};
ListaBTp[0.7] = {};
ListaBTp[0.8] = {};
ListaBTp[0.9] = {};

```

```

(* Este grupo esta llenando las listas BT para cada fracion transformada*)
Do[
  Clear[z];
  ListaprovBT[q] = {};
  Do[z := ListaFTVST[q][[i, 2]];
    If[z < 1, If[z >= 0.9, ListaprovBT[q] = AppendTo[ListaprovBT[q], ListaFTVST[q][[i]]]],
    {i, 1, Length[ListaFTVST[q]]}; Lo = Sort[ListaprovBT[q], #1[[2]] < #2[[2]] &];
    menorL = Lo[[1]]; ListaBTp[0.9] = AppendTo[ListaBTp[0.9], {menorL, q}], {q, 5, 30, 5}]
  Do[ListaBT[0.9] = AppendTo[ListaBT[0.9],
    {-ListaBTp[0.9][[i, 2]], ListaBTp[0.9][[i, 1, 1]]}], {i, 1, Length[ListaBTp[0.9]]};

Do[
  Clear[z];
  ListaprovBT[q] = {};
  Do[z := ListaFTVST[q][[i, 2]];
    If[z < 0.9, If[z >= 0.8, ListaprovBT[q] = AppendTo[ListaprovBT[q], ListaFTVST[q][[i]]]],
    {i, 1, Length[ListaFTVST[q]]}; Lo = Sort[ListaprovBT[q], #1[[2]] < #2[[2]] &];
    menorL = Lo[[1]]; ListaBTp[0.8] = AppendTo[ListaBTp[0.8], {menorL, q}], {q, 5, 30, 5}]
  Do[ListaBT[0.8] = AppendTo[ListaBT[0.8],

```

```

        {-ListaBTp[0.8][[i, 2]], ListaBTp[0.8][[i, 1, 1]]}, {i, 1, Length[ListaBTp[0.8]]}];

Do[
  Clear[z];
  ListaprovBT[q] = {};
  Do[z := ListaFTVST[q][[i, 2]];
    If[z < 0.8, If[z >= 0.7, ListaprovBT[q] = AppendTo[ListaprovBT[q], ListaFTVST[q][[i]]]],
      {i, 1, Length[ListaFTVST[q]]}; Lo = Sort[ListaprovBT[q], #1[[2]] < #2[[2]] &];
    menorL = Lo[[1]]; ListaBTp[0.7] = AppendTo[ListaBTp[0.7], {menorL, q}], {q, 5, 30, 5}]
Do[ListaBT[0.7] = AppendTo[ListaBT[0.7],
  {-ListaBTp[0.7][[i, 2]], ListaBTp[0.7][[i, 1, 1]]}, {i, 1, Length[ListaBTp[0.7]]}];

Do[
  Clear[z];
  ListaprovBT[q] = {};
  Do[z := ListaFTVST[q][[i, 2]];
    If[z < 0.7, If[z >= 0.6, ListaprovBT[q] = AppendTo[ListaprovBT[q], ListaFTVST[q][[i]]]],
      {i, 1, Length[ListaFTVST[q]]}; Lo = Sort[ListaprovBT[q], #1[[2]] < #2[[2]] &];
    menorL = Lo[[1]]; ListaBTp[0.6] = AppendTo[ListaBTp[0.6], {menorL, q}], {q, 5, 30, 5}]
Do[ListaBT[0.6] = AppendTo[ListaBT[0.6],
  {-ListaBTp[0.6][[i, 2]], ListaBTp[0.6][[i, 1, 1]]}, {i, 1, Length[ListaBTp[0.6]]}];

Do[
  Clear[z];
  ListaprovBT[q] = {};
  Do[z := ListaFTVST[q][[i, 2]];
    If[z < 0.6, If[z >= 0.5, ListaprovBT[q] = AppendTo[ListaprovBT[q], ListaFTVST[q][[i]]]],
      {i, 1, Length[ListaFTVST[q]]}; Lo = Sort[ListaprovBT[q], #1[[2]] < #2[[2]] &];
    menorL = Lo[[1]]; ListaBTp[0.5] = AppendTo[ListaBTp[0.5], {menorL, q}], {q, 5, 30, 5}]
Do[ListaBT[0.5] = AppendTo[ListaBT[0.5],
  {-ListaBTp[0.5][[i, 2]], ListaBTp[0.5][[i, 1, 1]]}, {i, 1, Length[ListaBTp[0.5]]}];

Do[
  Clear[z];
  ListaprovBT[q] = {};
  Do[z := ListaFTVST[q][[i, 2]];
    If[z < 0.5, If[z >= 0.4, ListaprovBT[q] = AppendTo[ListaprovBT[q], ListaFTVST[q][[i]]]],
      {i, 1, Length[ListaFTVST[q]]}; Lo = Sort[ListaprovBT[q], #1[[2]] < #2[[2]] &];
    menorL = Lo[[1]]; ListaBTp[0.4] = AppendTo[ListaBTp[0.4], {menorL, q}], {q, 5, 30, 5}]
Do[ListaBT[0.4] = AppendTo[ListaBT[0.4],
  {-ListaBTp[0.4][[i, 2]], ListaBTp[0.4][[i, 1, 1]]}, {i, 1, Length[ListaBTp[0.4]]}];

Do[
  Clear[z];
  ListaprovBT[q] = {};
  Do[z := ListaFTVST[q][[i, 2]];

```

```

If[z < 0.4, If[z >= 0.3, ListaprovBT[q] = AppendTo[ListaprovBT[q], ListaFTVST[q][[i]]]],
{i, 1, Length[ListaFTVST[q]]}; Lo = Sort[ListaprovBT[q], #1[[2]] < #2[[2]] &];
menorL = Lo[[1]]; ListaBTp[0.3] = AppendTo[ListaBTp[0.3], {menorL, q}], {q, 5, 30, 5}
Do[ListaBT[0.3] = AppendTo[ListaBT[0.3],
{-ListaBTp[0.3][[i, 2]], ListaBTp[0.3][[i, 1, 1]]}], {i, 1, Length[ListaBTp[0.3]]};

Do[
Clear[z];
ListaprovBT[q] = {};
Do[z := ListaFTVST[q][[i, 2]];
If[z < 0.3, If[z >= 0.2, ListaprovBT[q] = AppendTo[ListaprovBT[q], ListaFTVST[q][[i]]]],
{i, 1, Length[ListaFTVST[q]]}; Lo = Sort[ListaprovBT[q], #1[[2]] < #2[[2]] &];
menorL = Lo[[1]]; ListaBTp[0.2] = AppendTo[ListaBTp[0.2], {menorL, q}], {q, 5, 30, 5}
Do[ListaBT[0.2] = AppendTo[ListaBT[0.2],
{-ListaBTp[0.2][[i, 2]], ListaBTp[0.2][[i, 1, 1]]}], {i, 1, Length[ListaBTp[0.2]]};

Do[
Clear[z];
ListaprovBT[q] = {};
Do[z := ListaFTVST[q][[i, 2]];
If[z < 0.2, If[z >= 0.1, ListaprovBT[q] = AppendTo[ListaprovBT[q], ListaFTVST[q][[i]]]],
{i, 1, Length[ListaFTVST[q]]}; Lo = Sort[ListaprovBT[q], #1[[2]] < #2[[2]] &];
menorL = Lo[[1]]; ListaBTp[0.1] = AppendTo[ListaBTp[0.1], {menorL, q}], {q, 5, 30, 5}
Do[ListaBT[0.1] = AppendTo[ListaBT[0.1],
{-ListaBTp[0.1][[i, 2]], ListaBTp[0.1][[i, 1, 1]]}], {i, 1, Length[ListaBTp[0.1]]};

ListPlot[{ListaBT[0.9], ListaBT[0.8], ListaBT[0.7], ListaBT[0.6], ListaBT[0.5],
ListaBT[0.4], ListaBT[0.3], ListaBT[0.2], ListaBT[0.1]}, PlotStyle -> PointSize[0.01],
AxesLabel -> {" $\beta$ , ( °K min-1 )", "T, ( K ) "}, Joined -> True, PlotMarkers -> Automatic]
ListamemofqT = {};
Do[ListamemofqT = AppendTo[ListamemofqT, ListaBT[val $\xi$ ]], {val $\xi$ , 0.1, 0.6, 0.1}
Do[ListamemofqT = AppendTo[ListamemofqT, ListaBT[val $\xi$ ]], {val $\xi$ , 0.8, 0.9, 0.1}
ListamemofqT >> "D:\\Jorge\\Tesis\\noisotermico\\modelar los
parametros cineticos\\modelandoAISII1050farjas\\QvsTPafijPro1.nb"

```

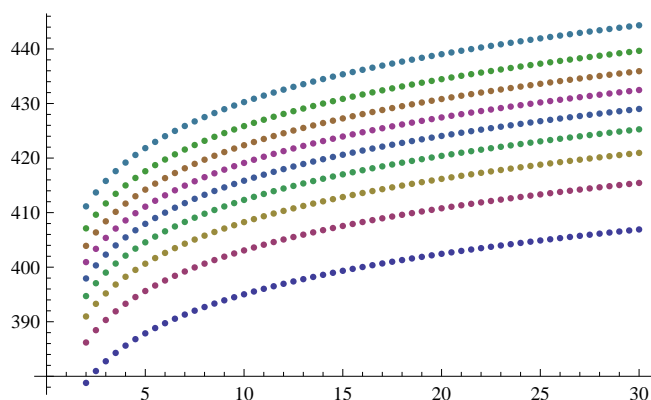
```

ListagfqT[0.7] = {{2., 403.906}, {2.5, 406.37}, {3., 408.40500000000003}, {3.5, 410.141},
  {4., 411.657}, {4.5, 413.003}, {5., 414.214}, {5.5, 415.315}, {6., 416.326},
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ListagfqT[0.6] = ListamemofqTr[[6]];
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ListagfqT[0.9] = ListamemofqTr[[8]];
ListagfqT[0.7] = ListamemofqTr[[9]];
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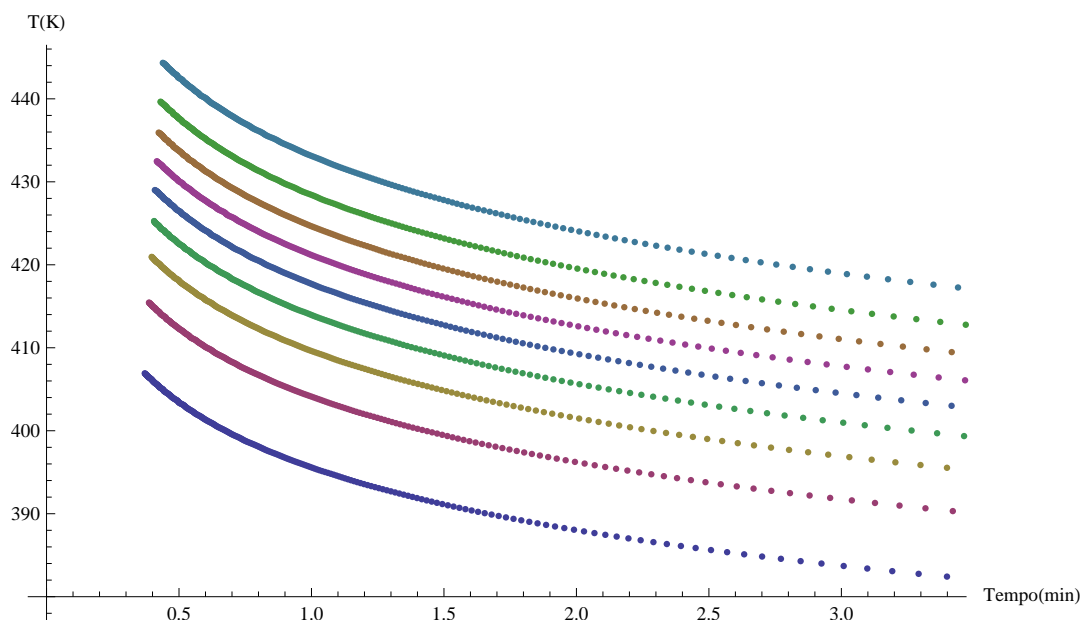
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  {valξ, 0.8, 0.9, 0.1}]
funTenq[0.7] = Interpolation[ListagfqT[0.7], Method → "Spline"];

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Do[ListagfTtiem[valξ] = {};
  Do[ListagfTtiem[valξ] =
    AppendTo[ListagfTtiem[valξ], {funTenq[valξ] ' [x], funTenq[valξ][x]}],
    {x, funTenq[valξ][[1, 1, 1]], funTenq[valξ][[1, 1, 2]], 0.1}], {valξ, 0.8, 0.9, 0.1}]

ListagfTtiem[0.7] = {};
Do[ListagfTtiem[0.7] =
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  {x, funTenq[0.7][[1, 1, 1]], funTenq[0.7][[1, 1, 2]], 0.1}]
ListPlot[{ListagfTtiem[0.1], ListagfTtiem[0.2], ListagfTtiem[0.3],
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  ListagfTtiem[0.8], ListagfTtiem[0.9]}, AxesLabel → {"Tempo(min)", "T(K)"}]
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```



## Anexo 16 Método A1

Considerando que  $P_0 = (\delta/l)_0$  e  $P_1 = (\delta/l)_1$  dependem da temperatura, como é mostrado na fig (3.2) e diferenciando a equação (3.1) com relação a  $T$  obtêm-se :

$$\begin{aligned} \frac{d\xi}{dT} &= \frac{d}{dT} [(P - P_0)(P_1 - P_0)^{-1}] = \\ &= \frac{d(P - P_0)}{dT} (P_1 - P_0)^{-1} + (P - P_0) \frac{d}{dT} (P_1 - P_0)^{-1} = \\ &= \left\{ \frac{dP}{dT} - \frac{dP_0}{dT} \right\} (P_1 - P_0)^{-1} + (P - P_0) \left\{ -1(P_1 - P_0)^{-2} \left( \frac{dP_1}{dT} - \frac{dP_0}{dT} \right) \right\} \Rightarrow \\ \frac{d\xi}{dT} &= \frac{1}{P_1 - P_0} \left\{ \left( \frac{dP}{dT} - \frac{dP_0}{dT} \right) - \frac{P - P_0}{P_1 - P_0} \left( \frac{dP_1}{dT} - \frac{dP_0}{dT} \right) \right\} \end{aligned} \quad (3.7)$$

E diferenciando novamente a equação (3.7) respeito a  $T$  teremos (Anexo 2, eq. (A.2.6)):

$$\frac{d^2\xi}{dT^2} = \frac{1}{P_1 - P_0} \left\{ \frac{d^2P}{dT^2} - \frac{2}{P_1 - P_0} \left( \frac{dP_1}{dT} - \frac{dP_0}{dT} \right) \left( \frac{dP}{dT} - \frac{dP_0}{dT} \right) + \frac{2}{(P_1 - P_0)^2} \left( \frac{dP_1}{dT} - \frac{dP_0}{dT} \right)^2 (P - P_0) \right\} \quad (3.8)$$

Quando  $T=T_i$  ;  $\left| \frac{dP}{dT} \right|_{T_i}$  é máximo e  $P$  não é muito grande pois  $P \in [P_0, P_1]$  e por esta

razão  $\left| \left( \frac{dP_1}{dT} - \frac{dP_0}{dT} \right) \left( \frac{P - P_0}{P_1 - P_0} \right) \right|_{T_i} \ll \left| \frac{dP}{dT} - \frac{dP_0}{dT} \right|_{T_i}$  o ultimo termino nas equações (7) e (8)

podem ser desprezados. Além disso em  $T=T_i$  ,  $\left( \frac{d^2P}{dT^2} \right)_{T_i} = 0$  , então a equação ( 3.8) se

transforma em :

$$\left( \frac{d^2\xi}{dT^2} \right)_{T_i} = - \frac{2}{(P_1 - P_0)^2} \left( \frac{dP_1}{dT} - \frac{dP_0}{dT} \right) \left( \frac{dP}{dT} - \frac{dP_0}{dT} \right) = - \frac{2}{P_1 - P_0} \left( \frac{dP_1}{dT} - \frac{dP_0}{dT} \right) \left( \frac{d\xi}{dT} \right)_{T_i}$$

Que pode ser escrita como:

$$\left( \frac{d^2\xi}{dT^2} \right)_{T_i} = -C(T) \left( \frac{d\xi}{dT} \right)_{T_i} \quad \text{onde:} \quad C(T) = \frac{2}{P_1 - P_0} \left( \frac{dP_1}{dT} - \frac{dP_0}{dT} \right) \quad (3.9)$$

Lembrando que:  $\beta = \frac{dT}{dt}$ ; podemos escrever:

$$\frac{d\xi}{dT} = \beta^{-1} \frac{d\xi}{dt} \quad \text{e} \quad \frac{d^2\xi}{dT^2} = \beta^{-2} \frac{d^2\xi}{dt^2} \quad (3.10)$$

Usando agora a equação (3.4):  $\xi = 1 - \exp(-\theta)^n$

$$\frac{d\xi}{dt} = \exp(-\theta^n) \theta^{n-1} n \frac{d\theta}{dt} \quad \text{e}$$

$$\frac{d^2\xi}{dt^2} = \left(\frac{d\theta}{dt}\right)^2 \exp(-\theta^n) \{-n^2 \theta^{2n-2} + n(n-1) \theta^{n-2}\} + n \theta^{n-1} \exp(-\theta^n) \frac{d^2\theta}{dt^2} \quad (3.11).$$

Substituindo (11) em (10) e o resultado em (9), obtém-se a equação diferencial:

$$\left(\theta \frac{d^2\theta}{dT^2}\right)_{Ti} - [n(\theta^n - 1) + 1] \left(\frac{d\theta}{dT}\right)_{Ti}^2 = -C(T) \theta \left(\frac{d\theta}{dT}\right)_{Ti} \quad (3.12)$$

Como  $n$  está compreendido no intervalo (1,4), e como na faixa de trabalho onde  $\xi$ , sofreu câmbios significativos (perto do ponto de inflexão, onde  $(d\xi/dT)$  é máxima),  $\theta \approx 1$ , então é possível fazer a seguinte aproximação para  $\theta^n = \exp(n \ln(\theta))$  que pode ser desenvolvido na série de potências:

$$\exp(u) = 1 + \frac{u}{1!} + \frac{u^2}{2!} + \frac{u^3}{3!} + \dots \Rightarrow \theta^n = \exp(n \ln(\theta)) = 1 + \frac{n \ln \theta}{1!} + \frac{(n \ln \theta)^2}{2!} + \frac{(n \ln \theta)^3}{3!} + \dots$$

Fazendo uso desta aproximação e substituindo a equação (3.6) em a equação (3.12) e como  $RT/E \ll 1$ , desprezando as potências superiores de  $RT/E$ , depois de algumas simplificações (Ver Anexo 3) se obtém [17] [48]:

$$\ln \frac{Ti^2}{\beta} = \ln \frac{E}{RK_0} + \frac{E}{RT} + RES_1 + RES_2 \quad (3.13)$$

$$\text{Onde:} \quad RES_1 = \frac{Q}{n^2} \frac{RTi^2}{E}, \quad RES_2 = 2 \frac{RTi}{E} * \left\{ 1 - \frac{1}{n^2} + n \ln \left[ \frac{Ti^2 RK(Ti)}{\beta E} \right] \right\} \text{ e}$$

$$Q(Ti) = 2 \left( \frac{\frac{dP_1}{dT} - \frac{dP_0}{dT}}{P_1 - P_0} \right)_{Ti}$$

Para uma transformação que envolve nucleação e crescimento e assumindo que os núcleos da nova fase ficam aleatoriamente distribuídos, Avrami, M [22,69,70], obteve a equação:

$$\xi = 1 - \exp[-v_{ext}] \quad (3.14)$$

Onde  $V_{ext}$  é o volume total por unidade de volume no tempo  $t$  desprezando a sobreposição dos grãos em crescimento:

$$v_{ext} = \int_0^t N(\tau) v(t, \tau) d\tau \quad (3.15)$$

Onde  $N(\tau)$  é a velocidade da nucleação por unidade de volume e  $v(t, \tau)$  é o volume transformado por um único núcleo, que há nascido em o tempo  $\tau$ , durante o tempo  $t$ :

$$v(z, t) = \sigma \left( \int_{\tau}^t G(z) dz \right)^m \quad (3.16)$$

Onde  $\sigma$  é um fator de forma (por exemplo,  $\sigma = 4/3 \pi$  para um grão esférico),  $G(z)$  é a velocidade de crescimento e  $m$  depende do mecanismo de crescimento (por exemplo,  $m = 3$  para um crescimento em três dimensões).

Na maior parte das situações praticas é possível assumir uma dependência do tipo de Arrhenius da temperatura para  $N(z)$  e  $G(z)$  [1]:

$$N = N_0 \exp\left[-\frac{EN}{k_b T}\right] \quad \text{e} \quad G = G_0 \exp\left[-\frac{EG}{k_b T}\right] \quad (3.17)$$

Só é possível escrever a equação de velocidade de uma reação na forma da equação (3.2) para o caso iso-cinético, o que é equivalente quando os parâmetros cinéticos não dependem do tempo ou da temperatura. De um modo simultâneo a mesma equação pode ser aplicada a uma reação não isotérmica quando a taxa de transformação depende só da temperatura e não da história térmica [71,72,73]. Esta condição é satisfeita especialmente na saturação de sítios para a nucleação, quando a nucleação ocorre antes do crescimento dos cristais [11] ou para a situação iso-cinética, especialmente quando  $N$  e  $G$  possuem a mesma energia de ativação [55].



Para uma velocidade de aquecimento constante  $\beta = \frac{dT}{dt} = cte$  e substituindo a equação (3.17) na equação (3.16):

$$v(\tau, t) = \sigma \left( \int_{\tau}^t G(z) dz \right)^m = \sigma \left( \int_{\tau}^t G_0 \exp \left( -\frac{E}{KbT} \right) dz \right)^m \quad (3.18)$$

E como  $\beta = \frac{dT}{dz} = cte \Rightarrow T = T_0 + \beta z$ , se usamos esta relação na equação (3.18) :

$$v(\tau, t) = \frac{\sigma G_0^m}{\beta^m} \left( \int_{T_0}^T \exp \left( -\frac{E}{KbT} \right) dT \right)^m \quad (3.19)$$

E esta integral é a mesma que aparece em o anexo 1 (A.1.3) , tomando o resultado (A.1.9):

$$v(\tau, t) = \frac{\sigma G_0^m}{\beta^m} \left\{ T \int_1^{\infty} \frac{\exp \left( -\frac{E}{KbT\mu} \right)}{\mu^2} d\mu - T_0 \int_1^{\infty} \frac{\exp \left( -\frac{E}{KbT\mu} \right)}{\mu^2} d\mu \right\}^m \quad (3.20)$$

Depois de multiplicar e dividir cada somando na eq. (3.20) por K/E:

$$v(\tau, t) = \sigma \left( \frac{GoE}{K\beta} \right)^m \left( \frac{\int_1^{\infty} \frac{\exp \left( -\frac{E}{KbT\mu} \right)}{\mu^2} d\mu}{\frac{E}{KbT}} - \frac{\int_1^{\infty} \frac{\exp \left( -\frac{E}{KbT\mu} \right)}{\mu^2} d\mu}{\frac{E}{KbT_0}} \right)^m \quad (3.21)$$

Usando a definição (A.1.7):

$$E_2(x) = \int_1^{\infty} \frac{\exp(-xy)}{y^2} dy \quad \text{y tomando } x = \frac{E}{KbT} = \frac{E}{RT}, \text{ lembrando que } R = NaKb \text{ onde}$$

Na é o numero de Avogadro e R a constante universal dos gases. Agora as unidades de E são ( KJ mol<sup>-1</sup>).

$$v(\tau, t) = \sigma \left( \frac{GoE}{K\beta} \right)^m \left( \frac{E_2(x)}{x} - \frac{E_2(x_0)}{x_0} \right)^m \quad (3.22)$$

Definindo :

$$P(x) = \frac{E_2(x)}{x} = \int_x^{\infty} \frac{\exp(-u)}{u^2} du \quad (3.23)$$

O volume transformado de um só núcleo, equação (3.22), se pode escrever:

$$v(\tau, t) = \sigma \left( \frac{GoE}{K\beta} \right)^m \left[ P \left( \frac{Eg}{RT} \right) - P \left( \frac{Eg}{RT_0} \right) \right]^m \quad (3.24)$$

E a fração de volume estendido,  $V_{ext}$ , pode ser calculado substituindo as equações (3.17) e (3.24) na equação (3.15).

$$V_{ext} = \int_0^t N(\tau) v(z, \tau) d\tau = \int_0^t No \exp\left(-\frac{En}{RT'}\right) \sigma\left(\frac{GoE}{K\beta}\right)^m \left[P\left(\frac{Eg}{RT}\right) - P\left(\frac{Eg}{RT'}\right)\right]^m d\tau \quad (3.25)$$

E como  $\beta = \frac{dT'}{d\tau} = cte \Rightarrow T' = To + \beta\tau, dT' = \beta d\tau$  e alem  $y = \frac{Eg}{RT}$  ;  $w = \frac{Eg}{RT'} \Rightarrow$

$$dw = -\frac{Eg}{R^2 T'^2} dT' \quad (3.26).$$

Substituindo (3.26) na equação (3.25) se obtém:

$$V_{ext} = -\frac{No\sigma Go^m}{\beta} \left(\frac{Eg}{\beta R}\right)^m \int_0^t \exp\left(-\frac{En}{RT'} \frac{Eg}{Eg}\right) (P(y) - P(w))^m \frac{RT'^2}{Eg} dw \quad (3.27)$$

Multiplicando e dividindo a Eq. (3.27) por  $Eg/R$ .

$$V_{ext} = -\frac{No\sigma Go^m}{\beta} \left(\frac{Eg}{\beta R}\right)^m \frac{Eg}{R} \int_0^t \exp\left(-\frac{En}{RT'} \frac{Eg}{Eg}\right) (P(y) - P(w))^m \frac{dw}{\frac{Eg^2}{R^2 T'^2}} \quad (3.28)$$

$$V_{ext} = No\sigma Go^m \left(\frac{Eg}{R\beta}\right)^{m+1} \int_w^\infty \exp\left(-\frac{En}{Eg} w'\right) \frac{1}{w'^2} (P(y) - P(w'))^m dw' \quad (3.29)$$

Esta integral pode ser analiticamente resolvida se  $Em=Eg$ . Neste caso:

$$\frac{dP(w')}{dw'} = -\frac{\exp(-w')}{w'^2}$$

Então a equação (3.29) pode-se escrever como:

$$V_{ext} = No\sigma Go^m \left(\frac{Eg}{R\beta}\right)^{m+1} \int_w^\infty -dP(w') (P(y) - P(w'))^m$$

$$V_{ext} = \frac{No\sigma Go^m}{m+1} \left(\frac{Eg}{R\beta}\right)^{m+1} P(y)^{m+1} = \left[Ko \frac{E}{R\beta} P\left(\frac{E}{RT}\right)\right]^{m+1} \quad (3.30)$$

Assumindo que  $P(\infty) = 0$  e

$$Ko = \left(\frac{No\sigma Go^m}{m+1}\right)^{\frac{1}{m+1}} \quad (3.31)$$

Esta ultima relação ,intuitivamente se sugere que:

$$K = \left( \frac{\sigma N G^m}{m+1} \right)^{\frac{1}{m+1}} \quad (3.32)$$

Substituindo na eq.(32) a eq. (17) :

$$K = \left( \frac{N o \sigma G o^m}{m+1} \right)^{\frac{1}{m+1}} \exp \left( -\frac{E}{RT} \right) . \text{Onde } E = \frac{E n + m E g}{m+1} \quad (3.33)$$

Substituindo a equação (3.30) em a equação (3.14) :

$$\xi = 1 - \exp \left\{ - \left[ K o \frac{E}{R \beta} P \left( \frac{E}{RT} \right) \right]^{m+1} \right\} \quad ( 3.34 )$$