Pg 16 \Rightarrow

$$\int_{x_1}^{x_2} u(x, t_2) dx = \int_{x_1}^{x_2} u(x, t_1) dx + \int_{t_1}^{t_2} f(u(x_1, t)) dt - \int_{t_1}^{t_2} f(u(x_2, t)) dt$$

$$\text{w/ } x_1 = -M; x_2 = M; t_1 = 0; t_2 = 1$$

$$= \int_{-M}^M u(x, 1) dx = \int_{-M}^M u(x, 0) dx + \int_0^1 f(u(-M, t)) dt - \int_0^1 f(u(M, t)) dt$$

$$M \text{ large enough } \begin{aligned} u(-M, t) &\rightarrow u_L \\ u(M, t) &\rightarrow u_R \end{aligned}$$

$$\begin{aligned} \int_{-M}^M u(x, 1) dx &= \int_{-M}^M u(x, 0) dx + f(u_L) - f(u_R) \\ &= \int_{-M}^0 u_L + \int_0^M u_R + \dots \end{aligned}$$

$$= u_L(M) + u_R M + \dots$$

$$u(x, 1) = w(x)$$

$$= M(u_L + u_R)$$

let $x_1 = 0 ; x_2 = M ; t_1 = 0 ; t_2 = 1$

$$\int_0^M u(x,t) dx = \underbrace{\int_0^M u(x,0) dx}_{u_r M} + \underbrace{\int_0^1 f(u(0,t)) dt}_{F(u_l, u_r)} - \underbrace{\int_0^1 f(u(M,t)) dt}_{f(u_r)} \quad *$$

$$\therefore \int_0^M \hat{w}(\xi) d\xi = M u_r + F(u_l, u_r) - f(u_r)$$

let ~~$x_1 = 0 ; x_2 = M ; t_1 = 0 ; t_2 = 1$~~ $x_1 = -M ; x_2 = 0 ; t_1 = 0 ; t_2 = 1$

$$\Rightarrow \int_{-M}^0 u(x,t) dx = \underbrace{\int_{-M}^0 u(x,0) dx}_{u_l M} + \underbrace{\int_0^1 f(u(-M,t)) dt}_{f(u_l)} - \underbrace{\int_0^1 f(u(0,t)) dt}_{F(u_l, u_r)} \quad **$$

From 14.6

$$\int_{-M}^0 \hat{w} d\xi + \int_0^M \hat{w} d\xi = M(u_l + u_r) + f(u_l) - f(u_r)$$

Put in from last eq (eq **)

$$\Rightarrow u_l M + f(u_l) - F(u_l, u_r) = M(u_l + u_r) + f(u_l) - f(u_r) - \int_0^M \hat{w} d\xi$$

$$\Rightarrow F(u_e, u_r) = \cancel{f(u_e) + f(u_r)}$$

$$= f(u_r) - Mu_r + \int_0^M \hat{w} d\varepsilon$$

put in $\int_0^M \hat{w} d\varepsilon$ from eq * to get.

$$\int_{-M}^0 \hat{w} d\varepsilon + \cancel{u_r M} + F(u_e, u_r) - \cancel{f(u_r)} = M(u_e + \cancel{u_r}) + \underline{f(u_e)} - \cancel{f(u_r)}$$

$$\Rightarrow F(u_e, u_r) = f(u_e) + Mu_e - \int_{-M}^0 \hat{w} d\varepsilon$$

12.55

$$\eta(u_j^{n+1}) \leq \eta(u_j^n) - \frac{k}{h} [\tau(u_j^n) - \tau(u_{j-1}^n)] \quad \text{discrete entropy eq.}$$

Thus inference on sol. $\hat{w}(\xi)$ is it

$$\int_{-M}^M \eta(\hat{w}(\xi)) d\xi \leq M(\eta(u_e) + \eta(u_r)) + (\tau(u_e) - \tau(u_r))$$

integral form of entropy condition

$$\int_{x_1}^{x_2} \eta(u(x, t_2)) dx \leq \int_{x_1}^{x_2} \eta(u(x, t_1)) dx - \int_{t_1}^{t_2} \tau(u(x_2, t)) dt + \int_{t_1}^{t_2} \tau(u(x_1, t)) dt$$

Then let $x_1 = -M$; $x_2 = M$; $t_1 = 0$; $t_2 = 1$

$$\begin{aligned} \Rightarrow \int_{-M}^M \eta(u(x, 1)) dx &\leq \int_{-M}^M \eta(u(x, 0)) dx - \tau(u_r) + \tau(u_e) \\ &= \int_{-M}^0 \eta(u_e) dx + \int_0^M \eta(u_r) dx \\ &= M\eta(u_e) + M\eta(u_r) \end{aligned}$$

$$\int_{-M}^M \eta(u(x, 1)) dx \leq M(\eta(u_e) + \eta(u_r)) + \tau(u_e) - \tau(u_r)$$

As before let $x_1=0; x_2=M; t_1=0; t_2=1$
to get

$$\int_0^M \rho(\omega(\xi)) d\xi \leq M \rho(u_r) + \underbrace{\int_0^1 f(u(0,t)) dt}_{-f(u_r)} - \underbrace{\int_0^1 f(u(M,t)) dt}_{-f(u_r)}$$

$$\leq M \rho(u_r) + \bar{I}(u_r, u_r) - f(u_r)$$

† let $x_1=-M; x_2=0; t_1=0; t_2=1$ to get

$$\int_{-M}^0 \rho(\omega(\xi)) d\xi \leq \rho(u_l) M$$

All steps follow as before to get 2 expressions for the numerical entropy flux.

$$\bar{I}(u_l, u_r) = f(u_l) + M \rho(u_l) - \int_{-M}^0 \rho(\omega(\xi)) d\xi$$

or

$$\bar{I}(u_l, u_r) = f(u_r) - M \rho(u_r) + \int_0^M \rho(\omega(\xi)) d\xi.$$

~~The Easy~~ The Easy Hyperbolic eq satisfies

$$\int_{-M}^M \hat{w}(\xi) d\xi = M(u_l + u_r) + \hat{f}(u_l) - \hat{f}(u_r)$$

For M large enough - But so ~~we require~~ ~~we require~~ ~~we require~~

Sol
$$\int_{-M}^M w(\xi) d\xi = M(u_l + u_r) + f(u_l) - f(u_r)$$

The Approx sol to L $\Rightarrow \int_{-M}^M \hat{w}(\xi) d\xi = M() + f(u_l) - f(u_r)$

By 14.6

" $w = U(x, t)$ "

$\therefore f(u_l) - f(u_r) = \hat{f}(u_l) - \hat{f}(u_r)$

Then by 14.9

$$F(u_l, u_r) = f(u_r) - Mu_r + \int_0^M \hat{w}(\xi) d\xi$$

$f(u_l) = f(u_r) + \hat{f}(u_l) - \hat{f}(u_r)$ from 14.13.

then $F(u_l, u_r) = \underline{f(u_r)} + \underline{\hat{f}(u_l)} - \underline{\hat{f}(u_r)} + Mu_l - \int_{-M}^0 \hat{w}(\xi) d\xi$

$$F(u_e; u_r) = f(u_r) - \hat{f}(u_r) + \underbrace{\hat{f}(u_e) + Mu_e - \int_{-M}^0 \hat{w}(\xi) d\xi}_{\hat{F}(u_e, u_r)} \quad 2$$

$\hat{F}(u_e, u_r)$
By 14.8

But $\hat{F}(u_e, u_r) = \hat{f}(\hat{w}(0))$

$$\therefore F(u_e, u_r) = f(u_r) - \hat{f}(u_r) + \hat{f}(\hat{w}(0)) \quad (14.14)$$

Also. $f(u_r) = \hat{f}(u_r) - \hat{f}(u_e) + f(u_e)$ put in eq

14.9

$$F(u_r, u_r) = \hat{f}(u_r) - \hat{f}(u_e) + f(u_e) - Mu_r + \int_0^M \hat{w}(\xi) d\xi$$

$$= f(u_e) - \hat{f}(u_e) + \hat{f}(u_r) - Mu_r + \int_0^M \hat{w}(\xi) d\xi$$

$\hat{F}(u_e, u_r)$

"

$\hat{f}(\hat{w}(0))$

$$= f(u_e) - \hat{f}(u_e) + \hat{f}(\hat{w}(0))$$

4.6
 Approx \hat{w}) $\int_{-M}^M \hat{w}(\xi) d\xi = M(u_e + u_r) + \hat{f}(u_e) - \hat{f}(u_r)$

~~Eq. **~~

if $\hat{A}(u_r, u_r)(u_r - u_e) = f(u_r) - f(u_e)$

Claim \Rightarrow * is true
 if $\hat{f}(u) = \hat{A}u$ know

$$\hat{w}(\xi) = u_e + \sum_{\lambda_p < \xi} \alpha_p \hat{r}_p$$

$\therefore \int_{-M}^M \hat{w}(\xi) d\xi = u_e 2M + \int_{\xi=-M}^M \sum_{\lambda_p < \xi} \alpha_p \hat{r}_p d\xi$

Let

$$\lambda_1 < \lambda_2 < \lambda_3 < \dots < \lambda_n$$

$$+ \int_{\xi=\lambda_1}^{\lambda_2} + \int_{\lambda_2}^{\lambda_3} + \int_{\lambda_3}^{\lambda_4} + \dots + \int_{\lambda_{n-1}}^{\lambda_n} + \int_{\lambda_n}^M$$

$$+ \alpha_1 \hat{r}_1 (\cancel{\lambda_2} - \lambda_1) + (\alpha_1 \cancel{\lambda_1} + \alpha_2 \hat{r}_2) (\lambda_3 - \cancel{\lambda_2})$$

$$+ (\alpha_1 \hat{r}_1 + \alpha_2 \hat{r}_2 + \alpha_3 \hat{r}_3) (\lambda_4 - \cancel{\lambda_3}) + \dots$$

$$+ (\alpha_1 \hat{r}_1 + \alpha_2 \hat{r}_2 + \dots + \alpha_{n-1} \hat{r}_{n-1}) (\lambda_n - \cancel{\lambda_{n-1}})$$

$$+ (\alpha_1 \hat{r}_1 + \alpha_2 \hat{r}_2 + \dots + \alpha_{n-1} \hat{r}_{n-1} + \alpha_n \hat{r}_n) (M - \cancel{\lambda_n})$$

$$= \alpha_1 \hat{r}_1 \left[(\cancel{\lambda_2} - \lambda_1) + (\cancel{\lambda_3} - \cancel{\lambda_2}) + (\cancel{\lambda_4} - \cancel{\lambda_3}) + \dots + (M - \cancel{\lambda_n}) \right]$$

$(\cancel{\lambda_n} - \cancel{\lambda_{n-1}})$

$$+ \alpha_2 \hat{r}_2 \left[\cancel{\lambda_3} - \lambda_2 + \cancel{\lambda_4} - \cancel{\lambda_3} + \dots + (\cancel{\lambda_n} - \cancel{\lambda_{n-1}}) + (M - \cancel{\lambda_n}) \right]$$

+ ... +

$$+ \alpha_{n-1} \hat{r}_{n-1} \left[(\cancel{\lambda_n} - \cancel{\lambda_{n-1}}) + (M - \cancel{\lambda_n}) \right]$$

$$+ \alpha_n \hat{r}_n [M - \cancel{\lambda_n}]$$

$$= \alpha_1 \hat{r}_1 [M - \lambda_1] + \alpha_2 \hat{r}_2 [M - \lambda_2] + \dots$$

$$\alpha_{n-1} \hat{r}_{n-1} [M - \lambda_{n-1}] + \alpha_n \hat{r}_n [M - \lambda_n]$$

$$= M \sum_p \alpha_p \hat{r}_p - \sum_p \lambda_p \alpha_p \hat{r}_p$$

$$= M(u_r - u_e) - \hat{A}(u_r - u_e)$$

$$\int_{-M}^M \hat{w}(z) dz = 2Mu_e + Mu_r - Mu_e - \hat{A}(u_r - u_e)$$

$$= M(u_e + u_r) + \hat{A}(u_e - u_r)$$

$$\text{if } \hat{A}(u_e - u_r) = f(u_e) - f(u_r)$$

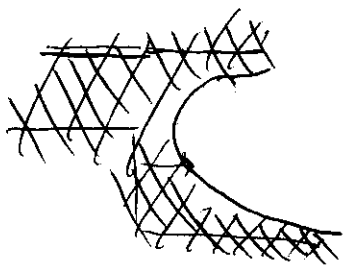
14.6 sets

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$$R.H. cond \Rightarrow f(u_r) - f(u_e) = S(u_r - u_e)$$

$$\sim 14.19: \Rightarrow A(u_r - u_e) = S(u_r - u_e)$$

$$\Rightarrow u_r - u_e \text{ e-vector of } A \text{ s.e. value } \lambda \therefore \underline{\hat{\sigma}(x,t)}$$



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$$F(u_e, u_r) = \hat{A} \hat{w}(0) + f(u_r) - \hat{A} u_r$$

$$= f(u_r) + \hat{A} \left[u_r - \sum_{\lambda > 0} \alpha_p \hat{r}_p \right] - \hat{A} u_r$$

$$= f(u_r) - \hat{A} \sum_{\lambda > 0} \alpha_p \hat{r}_p$$

$$= f(u_r) - \sum_{p=1}^m \hat{\lambda}_p^+ \alpha_p \hat{r}_p.$$

$$\text{or } F(u_e, u_r) = f(u_e) - \hat{A} u_e + \hat{A} \left[u_e + \sum_{\lambda < 0} \alpha_p \hat{r}_p \right]$$

$$= f(u_e) + \hat{A} \sum_{\lambda < 0} \alpha_p \hat{r}_p$$

$$= f(u_e) + \sum_{p=1}^m \hat{\lambda}_p^- \alpha_p \hat{r}_p.$$

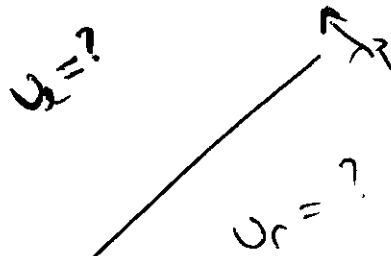
$$F(u_r, u_e) = \frac{1}{2}(f_l(u_e) + f_r(u_r)) + \frac{1}{2} \left(\sum \hat{\lambda}_p^- \alpha_p \hat{r}_p \leftarrow - \sum \hat{\lambda}_p^+ \alpha_p \hat{r}_p \right)^2$$

$$= \frac{1}{2}(f_l + f_r) + \frac{1}{2} \left(\sum \hat{\lambda}_p^+ \alpha_p \hat{r}_p - \sum \hat{\lambda}_p^- \dots \right)$$

$$- \frac{1}{2} \sum_{p=1}^m |\hat{\lambda}_p| \alpha_p \hat{r}_p$$

14.22 ^{Scalar case} $\Rightarrow f(u_e) + \hat{a}^-$

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$$u_{pr} = u_e + \sum_{i=1}^{p-1} \alpha_i \hat{r}_i$$

predicted



$$u_{pr} = u_{pe} + \alpha_i \hat{r}_i$$

By Roe.

$$(\lambda_{qr} - \lambda_{qe}) u_{qm} = (\lambda_q - \lambda_{qe}) u_{qe} + (\lambda_{qr} - \lambda_q) u_{qr}$$

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$$U_{qm} - U_{ql} = \begin{pmatrix} \hat{\lambda}_q - \lambda_{ql} & - \\ \lambda_{qr} - \lambda_{ql} & \frac{(\lambda_{qr} - \lambda_{ql})}{(\lambda_{qr} - \lambda_{ql})} \end{pmatrix} U_{ql} + \begin{pmatrix} \lambda_{qr} - \hat{\lambda}_r \\ \lambda_{qr} - \lambda_{ql} \end{pmatrix} U_{qr}$$

$$= \text{scribbled out}$$

$$= \frac{\hat{\lambda}_q - \lambda_{qr}}{\lambda_{qr} - \lambda_{ql}} U_{ql} + \frac{\lambda_{qr} - \hat{\lambda}_r}{\lambda_{qr} - \lambda_{ql}} U_{qr}$$

$$= \left(\frac{\lambda_{qr} - \hat{\lambda}_r}{\lambda_{qr} - \lambda_{ql}} \right) (U_{qr} - U_{ql})$$

$$\dagger \quad \cancel{U_{qr}} - U_{qm} = \frac{\lambda_{qr} - \lambda_{qe}}{\lambda_{qr} - \lambda_{qe}} \left(\alpha_q \hat{r}_q \right) \quad \text{from 14.26.}$$

$$U_{qr} - U_{qm} = - \left(\frac{\lambda_q - \lambda_{qe}}{\lambda_{qr} - \lambda_{qe}} \right) U_{qe} + \left(\frac{\lambda_{qr} - \lambda_q}{\lambda_{qr} - \lambda_{qe}} \right) U_{qr}$$

$$= - \left(\frac{\lambda_q - \lambda_{qe}}{\lambda_{qr} - \lambda_{qe}} \right) U_{qe} + \left(\frac{\lambda_q - \lambda_{qe}}{\lambda_{qr} - \lambda_{qe}} \right) U_{qr}$$

$$= \left(\frac{\lambda_q - \lambda_{qe}}{\lambda_{qr} - \lambda_{qe}} \right) (U_{qr} - U_{qe}) = \left(\frac{\lambda_q - \lambda_{qe}}{\lambda_{qr} - \lambda_{qe}} \right) (\alpha_q \hat{r}_q) \quad \text{from 14.26}$$

eq 14.34

$$\hat{\lambda}_{pe} = \lambda_{pe}^- \left(\frac{\lambda_{pr}^+ - \hat{\lambda}_p}{\lambda_{pr}^+ - \lambda_{pe}^-} \right)$$

$$\hat{\lambda}_{pr} = \lambda_{pr}^+ \left(\frac{\hat{\lambda}_p - \lambda_{pe}^-}{\lambda_{pr}^+ - \lambda_{pe}^-} \right)$$

P=9

$$\hat{\lambda}_{qe} = \lambda_{qe} \left(\frac{\lambda_{qr} - \hat{\lambda}_q}{\lambda_{qr} - \lambda_{qe}} \right)$$

eq 14.30 ✓

$$\hat{\lambda}_{qr} = \lambda_{qr} \left(\frac{\hat{\lambda}_q - \lambda_{qe}}{\lambda_{qr} - \lambda_{qe}} \right)$$

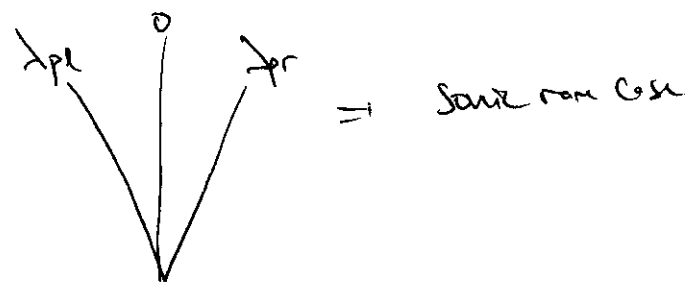
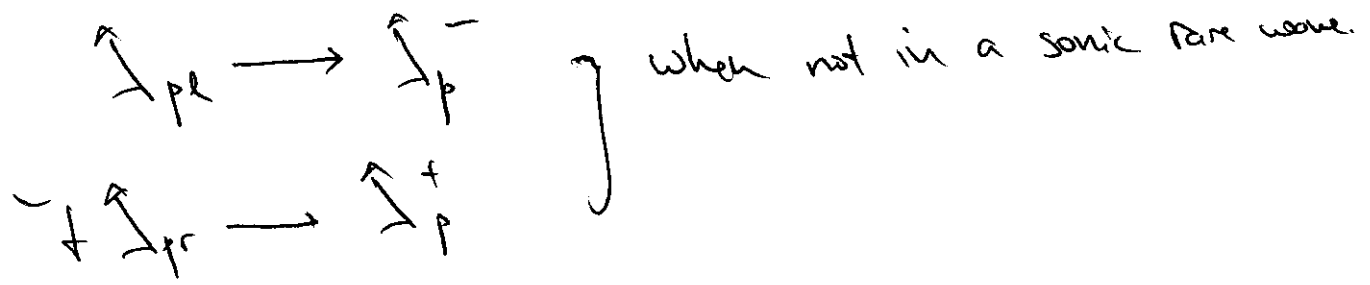
eq 14.31 ✓

P≠9

)

)

)



$0 < \lambda_{pe} < \lambda_{pr}$
 $\lambda_{pe}^- = 0$
 $\lambda_{pr}^+ = \lambda_{pr}$

$\therefore \lambda_{pe} = 0$

$\lambda_{pr} = \lambda_{pr} \left(\frac{\lambda_p - 0}{\lambda_{pr} - 0} \right) = \lambda_p \checkmark$

$\lambda_{pe} < \lambda_{pr} < 0$

$\lambda_{pe}^- = -\lambda_{pe}$

$\lambda_{pr}^+ = 0$

$\lambda_{pe} = -\lambda_{pe} \left(\frac{0 - \lambda_p}{0 + \lambda_{pe}} \right) = \lambda_p \checkmark$

$\lambda_{pr} = 0$

$$\hat{\Delta}_q = \hat{a} = (f(u_r) - f(u_l)) / (u_r - u_l)$$

$$\therefore 14.27 = u_m = \frac{(\hat{a} - f'(u_l))u_l + (f'(u_r) - \hat{a})u_r}{f'(u_r) - f'(u_l)}$$

$$= \frac{\hat{a}u_l - f'(u_l)u_l + f'(u_r)u_l}{f'(u_r) - f'(u_l)} + \frac{-f'(u_r)u_l + (f'(u_r) - \hat{a})u_r}{f'(u_r) - f'(u_l)}$$

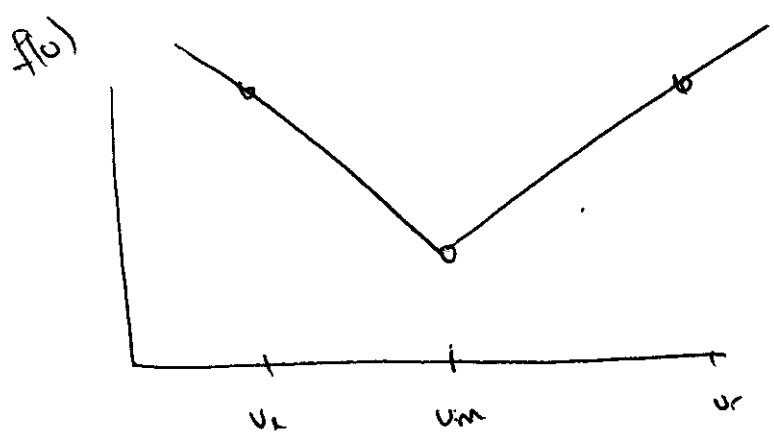
$$= u_l + \frac{\hat{a}u_l - f'(u_l)u_l + f'(u_r)u_l - f'(u_r)u_l + (f'(u_r) - \hat{a})u_r}{\Delta f'}$$

$$= u_l + \frac{(f'(u_r) - \hat{a})(u_r - u_l)}{\Delta f'}$$

$$)$$

$$\hat{\omega}(x/\epsilon) = \begin{cases} u_l & x/\epsilon < \hat{a} \\ u_r & x/\epsilon > \hat{a} \end{cases}$$

$$\Rightarrow \hat{\omega}(x/\epsilon) = \begin{cases} u_l & x/\epsilon < f'(u_l) \\ u_m & f'(u_l) < x/\epsilon < f'(u_r) \\ u_r & x/\epsilon > f'(u_r) \end{cases}$$



How show 14.41?

to get pt of intersection in 14.41.

$$f(u_e) + (u_m - u_e) f'(u_e) = f(u_r) + (u_m - u_r) f'(u_r)$$

$$\Rightarrow u_m = \frac{f(u_e) - f(u_r) + u_r f'(u_r) - u_e f'(u_e)}{-f'(u_e) + f'(u_r)}$$

$$= \frac{-u_e f'(u_e) + u_e f'(u_r) + (u_r - u_e) f'(u_r)}{-f'(u_e) + f'(u_r)}$$

$$= u_e + \frac{(f'(u_r) - \hat{a})(u_r - u_e)}{f'(u_r) - f'(u_e)} + \frac{f(u_e) - f(u_r)}{-f'(u_e) + f'(u_r)}$$

Think should be able to write as $u_r - \dots$

14.32

$$J(u, v_r) = f(u_e) + \sum_{p \neq q} \hat{\alpha}_p \hat{r}_p + \hat{\alpha}_q \hat{r}_q$$

$$\Rightarrow f(u_e) + \text{~~term~~} \quad (1) \quad \text{using def 14.34}$$

$$+ f'(u_e) \left(\frac{f'(u_r) - \hat{a}}{f'(u_r) - f'(u_e)} \right) (u_r - u_e) \quad (1)$$

14.14

$$J(u_e, u_r) = \text{~~scribbled out diagram~~}$$

$$= \hat{f}(\hat{\omega}(0)) + f(u_r) - \hat{f}(u_r)$$

$$= \hat{f}(u_m) + f(u_r) - \underbrace{\hat{f}(u_r)}_{f(u_r)}$$

$$= \hat{f}(u_m)$$

$$= f(u_e) + (u_m - u_e) f'(u_e)$$

$$\vec{v} = \begin{pmatrix} p \\ m \end{pmatrix} \quad f(u) = \begin{pmatrix} m \\ \frac{m^2}{p} + a^2 p \end{pmatrix} \quad \text{eqs 5.32 } \checkmark$$

$$f'(u) = \begin{pmatrix} 0 & 1 \\ -\frac{m^2}{p^2} + a^2 & \frac{2m}{p} \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ a^2 - v^2 & 2v \end{pmatrix} \quad \text{eq 14.48}$$

$$\vec{z} = p^{-1/2} \vec{v} = \begin{pmatrix} z_1 \\ z_2 \end{pmatrix} = \begin{pmatrix} p^{1/2} \\ m/p^{1/2} \end{pmatrix} = \begin{pmatrix} p^{1/2} \\ v p^{1/2} \end{pmatrix} \quad \text{eq 14.49}$$

$$\text{Then } \vec{u} = z_1 z_2 = \begin{pmatrix} z_1^2 \\ z_1 z_2 \end{pmatrix} \quad f(\vec{u}) = \begin{pmatrix} z_1 z_2 \\ z_2^2 + a^2 z_1^2 \end{pmatrix} \quad \text{eq 14.50}$$

$$\vec{z} = \frac{1}{2}(z_e + z_r) = \begin{pmatrix} \bar{z}_1 \\ \bar{z}_2 \end{pmatrix} = \frac{1}{2} \begin{pmatrix} \bar{p}_e^{1/2} + \bar{p}_r^{1/2} \\ \frac{m_e}{\bar{p}_e^{1/2}} + \frac{m_r}{\bar{p}_r^{1/2}} \end{pmatrix} \quad \text{eq 14.51}$$

$$\text{Then } u_e - u_r = \begin{pmatrix} z_{1e}^2 \\ z_{1e} z_{2e} \end{pmatrix} - \begin{pmatrix} z_{1r}^2 \\ z_{1r} z_{2r} \end{pmatrix} = \begin{pmatrix} z_{1e}^2 - z_{1r}^2 \\ z_{1e} z_{2e} - z_{1r} z_{2r} \end{pmatrix}$$

Cheating a bit (How do directly?) we notice that
 ↗ see notes that go w/ 8-14-01 4

$$\bar{z}_2(z_{1e} - z_{1r}) + \bar{z}_1(z_{2e} - z_{2r})$$

$$= \frac{1}{2}(z_{2e} + z_{2r})(z_{1e} - z_{1r}) + \frac{1}{2}(z_{1e} + z_{1r})(z_{2e} - z_{2r})$$

$$= \frac{1}{2} [z_{2e} z_{1e} - z_{2e} z_{1r} + z_{2r} z_{1e} - z_{2r} z_{1r}]$$

$$+\frac{1}{2} \left[z_{1e} z_{2e} - \cancel{z_{1e} z_{2r}} + \cancel{z_{1r} z_{2e}} - z_{1r} z_{2r} \right]$$

$$= z_{2e} z_{1e} - z_{1r} z_{2r}$$

Then

$$U_e - U_r = \begin{pmatrix} (z_{1e} - z_{1r})(z_{1e} + z_{1r}) \\ \bar{z}_2(z_{1e} - z_{1r}) + \bar{z}_1(z_{2e} - z_{2r}) \end{pmatrix} = \begin{pmatrix} 2\bar{z}_1(z_{1e} - z_{1r}) \\ \bar{z}_2(z_{1e} - z_{1r}) + \bar{z}_1(z_{2e} - z_{2r}) \end{pmatrix}$$

$$= \begin{pmatrix} 2\bar{z}_1 & 0 \\ \bar{z}_2 & \bar{z}_1 \end{pmatrix} \begin{pmatrix} z_{1e} - z_{1r} \\ z_{2e} - z_{2r} \end{pmatrix} = \begin{pmatrix} 2\bar{z}_1 & 0 \\ \bar{z}_2 & \bar{z}_1 \end{pmatrix} (z_e - z_r)$$

Also

$$f(U_e) - f(U_r) = \begin{pmatrix} z_{1e} z_{2e} - z_{1r} z_{2r} \\ a^2(z_{1e}^2 - z_{1r}^2) + z_{2e}^2 - z_{2r}^2 \end{pmatrix} \leftarrow \begin{array}{l} \text{same as last component} \\ \text{above} \end{array}$$

$$= \begin{pmatrix} \bar{z}_2(z_{1e} - z_{1r}) + \bar{z}_1(z_{2e} - z_{2r}) \\ a^2 2\bar{z}_1(z_{1e} - z_{1r}) + 2\bar{z}_2(z_{2e} - z_{2r}) \end{pmatrix}$$

$$= \begin{pmatrix} \bar{z}_2 & \bar{z}_1 \\ 2a^2 \bar{z}_1 & 2\bar{z}_2 \end{pmatrix} \begin{pmatrix} z_{1e} - z_{1r} \\ z_{2e} - z_{2r} \end{pmatrix}$$

$$[z] = \hat{B}^{-1}[u]$$

$$\therefore [f] = \hat{C}\hat{B}^{-1}[u]$$

$$\hat{A}(u_e, u_r) = \begin{bmatrix} \bar{z}_2 & \bar{z}_1 \\ 2a^2\bar{z}_1 & 2\bar{z}_2 \end{bmatrix} \frac{1}{2\bar{z}_1^2} \begin{bmatrix} \bar{z}_1 & 0 \\ -\bar{z}_2 & 2\bar{z}_1 \end{bmatrix}$$

$$= \frac{1}{2\bar{z}_1^2} \begin{bmatrix} 0 & 2\bar{z}_1^2 \\ 2a^2\bar{z}_1^2 - 2\bar{z}_2^2 & 2\bar{z}_1\bar{z}_2 \end{bmatrix}$$

$$= \begin{bmatrix} 0 & 1 \\ a^2 - \left(\frac{\bar{z}_2}{\bar{z}_1}\right)^2 & \left(\frac{\bar{z}_2}{\bar{z}_1}\right) \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ a^2 - \bar{v}^2 & \bar{v} \end{bmatrix} \quad \text{eq 14.53}$$

$$\omega \quad \bar{v} = \frac{\bar{z}_2}{\bar{z}_1} = \frac{P_e^{1/2} v_e + P_r^{1/2} v_r}{P_e^{1/2} + P_r^{1/2}} \quad \text{eq 14.54}$$

$$\omega \quad \hat{A} \text{ defined by } \hat{A}(u_e, u_r) = \begin{bmatrix} 0 & 1 \\ a^2 - \bar{v}^2 & \bar{v} \end{bmatrix}$$

General Method to determine Bode matrix terms :

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Given $z_{1e} z_{2e} - z_{1r} z_{2r}$ we want to write this in terms of $\bar{z}_1, \bar{z}_2, (z_{1e} - z_{1r}), (z_{2e} - z_{2r})$ This is equivalent to the following problem

$x_e y_e - x_r y_r$ written in terms of $\frac{1}{2}(x_e + x_r), \frac{1}{2}(y_e + y_r), x_e - x_r, y_e - y_r$.

Thus Notice that

$$\begin{aligned}x_e &= \frac{1}{2}(x_e + x_r) + \frac{1}{2}(x_e - x_r) \\y_e &= \frac{1}{2}(y_e + y_r) + \frac{1}{2}(y_e - y_r) \\x_r &= \frac{1}{2}(x_e + x_r) - \frac{1}{2}(x_e - x_r) \\y_r &= \frac{1}{2}(y_e + y_r) - \frac{1}{2}(y_e - y_r)\end{aligned}$$

Putting these

in the above expressions we get

$$\begin{aligned}x_e y_e - x_r y_r &= (\bar{x} + \frac{1}{2} \Delta x)(\bar{y} + \frac{1}{2} \Delta y) - (\bar{x} - \frac{1}{2} \Delta x)(\bar{y} - \frac{1}{2} \Delta y) \\&= \cancel{\bar{x}\bar{y}} + \frac{\bar{x}}{2} \Delta y + \frac{\bar{y}}{2} \Delta x + \frac{1}{4} \cancel{\Delta x \Delta y} \\&\quad - \cancel{\bar{x}\bar{y}} + \frac{\bar{x}}{2} \Delta y + \frac{\bar{y}}{2} \Delta x - \frac{1}{4} \cancel{\Delta x \Delta y} \\&= \bar{x} \Delta y + \bar{y} \Delta x\end{aligned}$$

Ex 14.4 eq 5.38 are

$$\begin{pmatrix} v \\ \phi \end{pmatrix}_t + \begin{pmatrix} v^2/2 + \phi \\ \phi v \end{pmatrix}_x = 0$$

$$\{\phi = gh\}$$

Then following the development of a Roe matrix for the isotherm eqs we note that a characteristic of the transformation done there is that the flux is rewritten in terms of products thus let

$$\vec{z} = \phi^{-1/2} \vec{u} = \begin{pmatrix} v/\phi^{1/2} \\ \phi^{1/2} \end{pmatrix} \equiv \begin{pmatrix} z_1 \\ z_2 \end{pmatrix}$$

Then $\vec{u} = \phi^{1/2} \vec{z} = z_2 \vec{z} = \begin{pmatrix} z_1 z_2 \\ z_2^2 \end{pmatrix}$ good two table products

Also $\vec{f}(\vec{u}) = \begin{pmatrix} v^2/2 + \phi \\ \phi v \end{pmatrix} = \begin{pmatrix} \frac{z_1^2 z_2^2}{2} + z_2^2 \\ z_2^2 z_1 z_2 \end{pmatrix}$

Not so good. Many products of high order maybe difficult to factor z_2 out of this expression.

Based on simple form of 14.55-14.56 we look at eigenvalues

$$J(\vec{u}) = \begin{pmatrix} v^2/2 + \phi \\ \phi v \end{pmatrix}$$

$$J'(\vec{u}) = \begin{pmatrix} v & 1 \\ \phi & v \end{pmatrix}$$

$$\lambda_{1,2} = v \pm \sqrt{\phi}$$

$$\begin{aligned} \text{Now } u_e - u_r &= \begin{pmatrix} z_e z_{e\bar{e}} - z_r z_{r\bar{r}} \\ z_{ze}^2 - z_{zr}^2 \end{pmatrix} = \begin{pmatrix} \bar{z}_2(z_e - z_r) + \bar{z}_1(z_{ze} - z_{zr}) \\ 2\bar{z}_2(z_e - z_r) \end{pmatrix} \\ &= \begin{pmatrix} \bar{z}_2 & \bar{z}_1 \\ 2\bar{z}_2 & 0 \end{pmatrix} \begin{pmatrix} z_e - z_r \\ z_{ze} - z_{zr} \end{pmatrix} = \begin{pmatrix} \bar{z}_2 & \bar{z}_1 \\ 2\bar{z}_2 & 0 \end{pmatrix} \begin{pmatrix} z_e - z_r \\ z_{ze} - z_{zr} \end{pmatrix} \end{aligned}$$

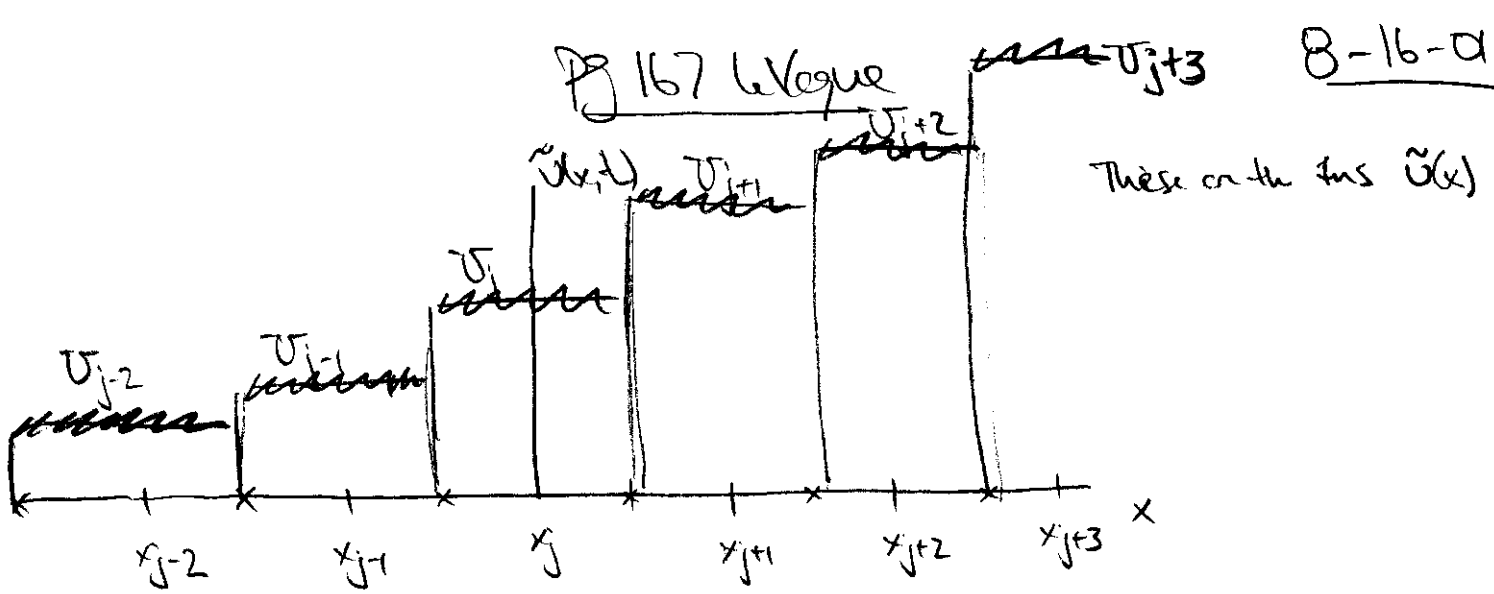
$$\begin{aligned} \text{Now } \|u_e\|^2 - \|u_r\|^2 &= \begin{pmatrix} \frac{z_{ze}^2 z_{ze}^2}{2} - \frac{z_{zr}^2 z_{zr}^2}{2} + z_{ze}^2 - z_{zr}^2 \\ z_{ze}^2 z_{ze} z_{ze} - z_{zr}^2 z_{zr} z_{zr} \end{pmatrix} \\ &= \begin{pmatrix} \frac{1}{2}(z_e z_{e\bar{e}} - z_r z_{r\bar{r}})(z_e z_{e\bar{e}} + z_r z_{r\bar{r}}) + 2\bar{z}_2(z_{ze} - z_{zr}) \\ z_{ze}^3 z_{ze} - z_{zr}^3 z_{zr} \end{pmatrix} \end{aligned}$$

$$* \quad z_e z_{e\bar{e}} - z_r z_{r\bar{r}} = \bar{z}_2(z_e - z_r) + \bar{z}_1(z_{ze} - z_{zr}) \text{ from previous part}$$

$\frac{1}{2} (z_{1e}^2 z_{2e}^2 - z_{1r}^2 z_{2r}^2)$ Following notes on pg 8-14-01 4

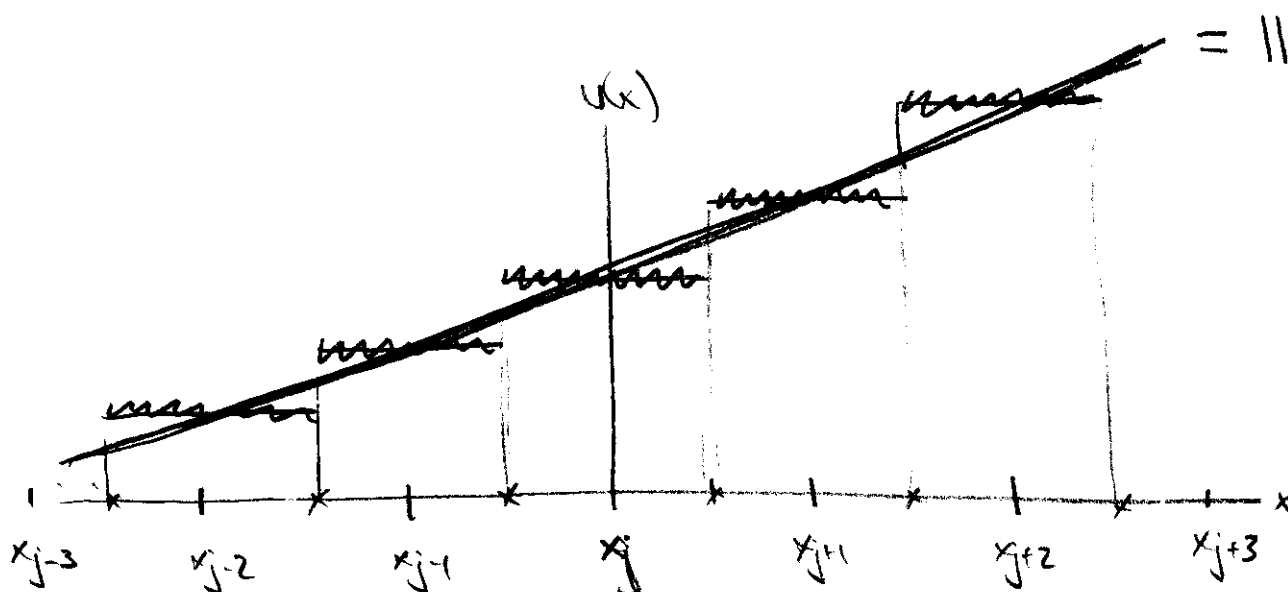
$$= \frac{1}{2} (x_e^2 y_e^2 - x_r^2 y_r^2) = \frac{1}{2} \left(\left(\bar{x} + \frac{1}{2} \Delta x \right)^2 \left(\bar{y} + \frac{1}{2} \Delta y \right)^2 - \left(\bar{x} - \frac{1}{2} \Delta x \right)^2 \left(\bar{y} - \frac{1}{2} \Delta y \right)^2 \right)$$

This does not simplify as required \Rightarrow perhaps our definition of \vec{z} is incorrect. What is a better choice for \vec{z} ?



$$\|u(x)\|_1 = \int_{-\infty}^{\infty} |u(x)| dx = \sum_{j=-\infty}^{+\infty} \int_{x_{j-1/2}}^{x_{j+1/2}} |u(x)| dx = \sum_{j=-\infty}^{\infty} |u_j| h = h \sum_{j=-\infty}^{\infty} |u_j|$$

= $\|u\|_1$ eq 15.25



Step fn is Restricted to constant cells given by Avg of

$$\|u\|_1 \equiv h \sum_{j=-\infty}^{+\infty} |u_j| = \sum_{j=-\infty}^{+\infty} h |u_j|$$

Now consider

$$\int_{x_j - h/2}^{x_j + h/2} |u(x)| dx \geq \int_{x_j - h/2}^{x_j + h/2} u(x) dx = h \frac{1}{h} \int_{x_j - h/2}^{x_j + h/2} u(x) dx$$

$$= h \bar{u}_j$$

Then since

$$\int_{x_j - h/2}^{x_j + h/2} |u(x)| dx \geq 0 \quad \left| \int_{x_j - h/2}^{x_j + h/2} u(x) dx \right| = \int_{x_j - h/2}^{x_j + h/2} |u(x)| dx$$

So that

$$h |\bar{u}_j| \leq \int_{x_j - h/2}^{x_j + h/2} |u(x)| dx \quad \text{Then}$$

$$\| \bar{u} \|_1 \leq \sum_{j=-\infty}^{+\infty} \int_{x_j - h/2}^{x_j + h/2} |u(x)| dx = \| u \|_1 \quad \text{eq 15.26}$$

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$$TV(u) = \sum_{j=-\infty}^{\infty} |u_{j+1}^n - u_j^n| = \sum_{j=-\infty}^{\infty} |u_j^n - u_{j-1}^n| = \sum_{j=-\infty}^{\infty} |u_j^n - v_j^n|$$

\parallel
 v_j^n

$$= \frac{1}{h} \|u - v\|_1$$

Ex 15.2 Lex-Friedrich 12.15 is

$$u_j^{n+1} = \frac{1}{2}(u_{j-1}^n + u_{j+1}^n) - \frac{k}{2h}(f(u_{j+1}^n) - f(u_{j-1}^n))$$

↓ for v_j^n is

$$v_j^{n+1} = \frac{1}{2}(v_{j-1}^n + v_{j+1}^n) - \frac{k}{2h}(f(v_{j+1}^n) - f(v_{j-1}^n))$$

Defining $w_j^n = u_j^n - v_j^n$ we get

$$w_j^{n+1} = \frac{1}{2}(w_{j-1}^n + w_{j+1}^n) - \frac{k}{2h} \left[(f(u_{j+1}^n) - f(v_{j+1}^n)) - (f(u_{j-1}^n) - f(v_{j-1}^n)) \right]$$

Apply by smoothness of f (mean value thm)

$$f(u_j^n) - f(v_j^n) = f'(\theta_j^n)(u_j^n - v_j^n) \quad \theta_j^n \text{ is between } u_j^n \text{ and } v_j^n$$

$$\Rightarrow w_j^{n+1} = \frac{1}{2}(w_{j-1}^n + w_{j+1}^n) - \frac{k}{2h} f'(\theta_{j+1}^n) w_{j+1}^n + \frac{k}{2h} f'(\theta_{j-1}^n) w_{j-1}^n$$

$$= \frac{1}{2} \left[1 + \frac{k}{h} f'(\theta_{j-1}^n) \right] w_{j-1}^n + \frac{1}{2} \left[1 - \frac{k}{h} f'(\theta_{j+1}^n) \right] w_{j+1}^n$$

$$-1 \leq \frac{k}{h} f'(\omega) \leq +1 \quad \forall \quad \min_j (\omega_j) \leq \omega \leq \max_j (\omega_j)$$

$$\begin{aligned} \therefore 0 \leq 1 + \frac{k}{h} f(\theta_{j+1}^n) \leq 2 &\Rightarrow 0 \leq \frac{1}{2} \left(1 + \frac{k}{h} f(\theta_{j+1}^n)\right) \leq 1 \\ + \quad 0 \leq 1 - \frac{k}{h} f(\theta_{j+1}^n) \leq 2 &\Rightarrow 0 \leq \frac{1}{2} \left(1 - \frac{k}{h} f(\theta_{j+1}^n)\right) \leq 1 \end{aligned}$$

So that

$$\begin{aligned} |W_j^{n+1}| &\leq \frac{1}{2} (1 + \alpha_{j+1}) |W_{j-1}^n| + \frac{1}{2} (1 - \alpha_{j+1}) |W_{j+1}^n| \\ &= \frac{1}{2} |W_{j-1}^n| + \frac{1}{2} |W_{j+1}^n| + \frac{\alpha_{j-1}}{2} |W_{j-1}^n| - \frac{\alpha_{j+1}}{2} |W_{j+1}^n| \end{aligned}$$

Multiplying by h & summing from $j = -\infty$ to $+\infty$ gives

$$h \sum_{j=-\infty}^{+\infty} |W_j^{n+1}| \leq h \sum_{j=-\infty}^{+\infty} |W_j^n| + 0$$

~~~~~

$$\|W^{n+1}\|_1 \leq \|W^n\|_1$$



6-10

$$u_j^{n+1} = u_j^n - \frac{\nu}{2}(u_{j+1}^n - u_{j-1}^n) + \frac{1}{2}\nu^2(u_{j+1}^n - 2u_j^n + u_{j-1}^n)$$

$$= u_j^n - \nu(u_j^n - u_{j-1}^n) + \nu u_j^n - \frac{\nu}{2}u_{j-1}^n - \frac{\nu}{2}u_{j+1}^n$$

$$+ \frac{1}{2}\nu^2(u_{j+1}^n - 2u_j^n + u_{j-1}^n)$$

$$= u_j^n - \nu(u_j^n - u_{j-1}^n) - \frac{\nu}{2}(u_{j+1}^n - 2u_j^n + u_{j-1}^n)$$

$$+ \frac{1}{2}\nu^2(\Delta_j^2 u_j^n)$$

$$= u_j^n - \nu(u_j^n - u_{j-1}^n) - \frac{1}{2}\nu(1-\nu)(u_{j+1}^n - 2u_j^n + u_{j-1}^n)$$

$$\nu = \frac{ak}{h}$$

$$u_j^{n+1} = u_j^n - \frac{k}{h} \left[ \left( au_j^n + \frac{1}{2}a(1-\nu)(u_{j+1}^n - u_j^n) \right) - \left( au_{j-1}^n + \frac{1}{2}a(1-\nu)(u_j^n - u_{j-1}^n) \right) \right]$$

$$\therefore F(u; j) = au_j + \frac{1}{2}a(1-\nu)(u_{j+1} - u_j)$$

w/o  $u_j^n$

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$$-v(u_j^n) + v u_{j-1}^n - \frac{1}{2} [v u_{j+1}^n - 2v u_j^n + v u_{j-1}^n - v^2 u_{j+1}^n + 2v^2 u_j^n - v^2 u_{j-1}^n]$$

$$= v \left( +1 - \frac{1}{2} + \frac{1}{2}v \right) u_{j-1}^n + v \left( \cancel{+1} + \cancel{+1} - v \right) u_j^n + v \left( \frac{1}{2} + \frac{v}{2} \right) u_{j+1}^n$$

$$= \frac{1}{2}v (u_{j+1}^n - u_{j-1}^n) + \frac{1}{2}v^2 (u_{j+1}^n - 2u_j^n + u_{j-1}^n) \quad \checkmark$$

Ex 16.1

$$F(U, j) = aU_j + \frac{1}{2} a(1-\nu)(U_{j+1} - U_j) \phi(\theta_j)$$

Then method is

$$U_j^{n+1} = U_j^n - \frac{k}{h} [F(U_{j-p}^n, U_{j-p+1}^n, \dots, U_{j+1}^n) - F(U_{j-p-1}^n, U_{j-p}^n, \dots, U_{j+1}^n)]$$

thus

$$U_j^{n+1} = U_j^n - \frac{k}{h} \left\{ aU_j^n + \frac{1}{2} a(1-\nu)(U_{j+1}^n - U_j^n) \phi(\theta_j) - aU_{j+1}^n - \frac{1}{2} a(1-\nu)(U_j^n - U_{j-1}^n) \phi(\theta_{j-1}) \right\}$$

$$\Rightarrow U_j^n + k(U_j^n)^n + \frac{k^2}{2}(U_j^n)^n + \frac{k^3}{6}(U_j^n)^n + O(k^4)$$

$$= U_j^n - \frac{k}{h} \left\{ aU_j^n + \frac{1}{2} a(1-\nu) \left( h(U_j^n)^n + \frac{h^2}{2}(U_{xx})_j^n + \frac{h^3}{6}(U_{xxx})_j^n + O(h^4) \right) \phi(\theta_j) \right.$$

$$\left. - a \left[ U_{j+1}^n - h(U_x)_j^n + \frac{h^2}{2}(U_{xx})_j^n - \frac{h^3}{6}(U_{xxx})_j^n + O(h^4) \right] \right\}$$

$$- \frac{1}{2} a(1-\nu) \left( U_j^n - \left( U_j^n - h(U_x)_j^n + \frac{h^2}{2}(U_{xx})_j^n - \frac{h^3}{6}(U_{xxx})_j^n + O(h^4) \right) \right) \phi(\theta_{j-1}) \left. \right\}$$

$$\Rightarrow (\sigma_t)_j^n + \frac{k}{2} (\sigma_{tt})_j^n + \frac{k^2}{6} (\sigma_{ttt})_j^n + \alpha k^3$$

$$= - \frac{1}{h} \int \left[ \frac{1}{2} a(1-\nu) \left( h(\sigma_x)_j^n + \frac{h^2}{2} (\sigma_{xx})_j^n + \frac{h^3}{6} (\sigma_{xxx})_j^n + \alpha h^4 \right) \phi(\theta_j) \right.$$

$$\left. + ah(\sigma_x)_j^n - \frac{ah^2}{2} (\sigma_{xx})_j^n + \frac{ah^3}{6} (\sigma_{xxx})_j^n + \alpha h^4 \right)$$

$$\left. + \frac{1}{2} a(1-\nu) \left( -h(\sigma_x)_j^n + \frac{h^2}{2} (\sigma_{xx})_j^n - \frac{h^3}{6} (\sigma_{xxx})_j^n + \alpha h^4 \right) \phi(\theta_{j-1}) \right] \quad \checkmark$$

$\Rightarrow$

$$(\sigma_t)_j^n + \frac{k}{2} (\sigma_{tt})_j^n + \frac{k^2}{6} (\sigma_{ttt})_j^n + \alpha k^3$$

$$= - \int \left[ \frac{1}{2} a(1-\nu) \left( (\sigma_x)_j^n + \frac{h}{2} (\sigma_{xx})_j^n + \frac{h^2}{6} (\sigma_{xxx})_j^n + \alpha h^3 \right) \phi(\theta_j) \right.$$

$$\left. + a(\sigma_x)_j^n - \frac{ah}{2} (\sigma_{xx})_j^n + \frac{ah^2}{6} (\sigma_{xxx})_j^n + \alpha h^3 \right)$$

$$\left. - \frac{1}{2} a(1-\nu) \left( (\sigma_x)_j^n - \frac{h}{2} (\sigma_{xx})_j^n + \frac{h^2}{6} (\sigma_{xxx})_j^n + \alpha h^3 \right) \phi(\theta_{j-1}) \right] \quad \checkmark$$

$$\Rightarrow (\sigma_t)_j^n + a(\sigma_x)_j^n = - \frac{k}{2} (\sigma_{tt})_j^n - \frac{k^2}{6} (\sigma_{ttt})_j^n + \alpha k^3$$

$$+ \frac{ah}{2} (\sigma_{xx})_j^n - \frac{ah^2}{6} (\sigma_{xxx})_j^n + \alpha h^3$$

$$-\frac{a(1-\nu)}{2} \left[ (u_x)_j^* (\phi(\theta_j) - \phi(\theta_{j-1})) + \frac{h}{2} (u_{xx})_j^* (\phi(\theta_j) + \phi(\theta_{j-1})) \right. \\ \left. + \frac{h^2}{6} (u_{xxx})_j^* (\phi(\theta_j) - \phi(\theta_{j-1})) + O(h^3) \right]$$

Now  $\phi(\theta_j) - \phi(\theta_{j-1})$

$$= \phi\left(\frac{\theta_j - \theta_{j-1}}{\theta_{j+1} - \theta_j}\right) - \phi\left(\frac{\theta_{j-1} - \theta_{j-2}}{\theta_j - \theta_{j-1}}\right)$$

So expanding  $\theta_j$  1st

$$\theta_j = \frac{\theta_j - \theta_{j-1}}{\theta_{j+1} - \theta_j} = \frac{\theta_j - (\theta_j - h(u_x)_j + \frac{h^2}{2}(u_{xx})_j + O(h^3))}{h(u_x)_j + \frac{h^2}{2}(u_{xx})_j + O(h^3)}$$

$$= \frac{h(u_x)_j - \frac{h^2}{2}(u_{xx})_j + O(h^3)}{h(u_x)_j + \frac{h^2}{2}(u_{xx})_j + O(h^3)}$$

$$= \frac{h(u_x)_j - \frac{h^2}{2}(u_{xx})_j + O(h^2)}{h(u_x)_j + \frac{h^2}{2}(u_{xx})_j + O(h^2)}$$

Assuming  $(u_x)_j \neq 0$

$$= \frac{1 - \frac{h}{2} \frac{(u_{xx})_j}{(u_x)_j} + O(h^2)}{1 + \frac{h}{2} \frac{(u_{xx})_j}{(u_x)_j} + O(h^2)}$$

$$\begin{aligned} \theta_j &\approx \left(1 - \frac{h}{2} \frac{(v_{xx})_j}{(v_x)_j} + \alpha h^2\right) \left(1 - \frac{h}{2} \frac{(v_{xx})_j}{(v_x)_j} + \alpha h^2\right) \\ &= 1 - \frac{h}{2} \frac{(v_{xx})_j}{(v_x)_j} - \frac{h}{2} \frac{(v_{xx})_j}{(v_x)_j} + \alpha h^2 \end{aligned}$$

Now:  $v_t + a v_x = 0$

$$v_{tt} + a(v_t)_x = 0$$

$$v_{tt} - a^2 v_{xx} = 0$$

Then  $\frac{-ka^2}{2} v_{xx} + \frac{ah}{2} v_{xx}$

$$(ka-h)v_{xx}$$

$$v = \frac{k}{h} a$$

Method:  $\sigma_j^{n+1} = \sigma_j^n - \frac{k}{h} (F(\sigma_j^n) - F(\sigma_{j-1}^n))$

$\alpha$  Flux  $F(\sigma_j^n) = a\sigma_j + \frac{1}{2}a(1-v)(\sigma_{j+1} - \sigma_j)\phi_j$

Then

$$\sigma_j^{n+1} = \sigma_j^n - \frac{k}{h} \left[ a\sigma_j + \frac{1}{2}a(1-v)(\sigma_{j+1} - \sigma_j)\phi_j - a\sigma_{j-1} - \frac{1}{2}a(1-v)(\sigma_j - \sigma_{j-1})\phi_{j-1} \right]$$

$$= \sigma_j^n - \frac{k}{h} \left[ a(\sigma_j - \sigma_{j-1}) + \frac{1}{2}a(1-v) \left[ (\sigma_{j+1} - \sigma_j)\phi_j - (\sigma_j - \sigma_{j-1})\phi_{j-1} \right] \right]$$

$$= \sigma_j^n - v(\sigma_j - \sigma_{j-1}) - \frac{1}{2}v(1-v) \left[ (\sigma_{j+1} - \sigma_j)\phi_j - (\sigma_j - \sigma_{j-1})\phi_{j-1} \right]$$

$$= \sigma_j^n + \left[ -v + \frac{v}{2}(1-v)\phi_{j-1} \right] (\sigma_j - \sigma_{j-1})$$

$$+ \left[ -\frac{1}{2}v(1-v)\phi_j \right] (\sigma_{j+1} - \sigma_j)$$

$$\Rightarrow \sigma_j^{n+1} = \sigma_j^n - (v - \frac{v}{2}(1-v)\phi_{j-1})(\sigma_j - \sigma_{j-1})$$

$$- \frac{1}{2}v(1-v)\phi_j(\sigma_{j+1} - \sigma_j) \quad \text{eq 16.17}$$

Consider

$$\sigma_{j+1}^{n+1} - \sigma_j^{n+1} = \sigma_{j+1} - G_j(\sigma_{j+1} - \sigma_j) + D_{j+1}(\sigma_{j+2} - \sigma_{j+1})$$

$$- \sigma_j + G_{j-1}(\sigma_j - \sigma_{j-1}) - D_j(\sigma_{j+1} - \sigma_j)$$



$$\begin{aligned}
 \therefore v_{j+1}^{n+1} - v_j^{n+1} &= C_{j-1}(\sigma_j - \sigma_{j-1}) \\
 &+ \sigma_{j+1} - \sigma_j - (c_j + D_j)(\sigma_{j+1} - \sigma_j) \\
 &+ D_{j+1}(\sigma_{j+2} - \sigma_{j+1}) \\
 &= (1 - c_j - D_j)(\sigma_{j+1} - \sigma_j) + D_{j+1}(\sigma_{j+2} - \sigma_{j+1}) \\
 &+ c_{j-1}(\sigma_j - \sigma_{j-1})
 \end{aligned}$$

$$\begin{aligned}
 |v_{j+1}^{n+1} - v_j^{n+1}| &\leq |1 - c_j - D_j| |\sigma_{j+1} - \sigma_j| + |D_{j+1}| |\sigma_{j+2} - \sigma_{j+1}| \\
 &+ |c_{j-1}| |\sigma_j - \sigma_{j-1}|
 \end{aligned}$$

Summing over  $j$

$$\begin{aligned}
 \sum_{j=-\infty}^{+\infty} |v_{j+1}^{n+1} - v_j^{n+1}| &\leq \sum_{j=-\infty}^{+\infty} |1 - c_j - D_j| |\sigma_{j+1} - \sigma_j| + \sum_{j=-\infty}^{+\infty} |D_{j+1}| |\sigma_{j+2} - \sigma_{j+1}| \\
 &+ \sum_{j=-\infty}^{+\infty} |c_{j-1}| |\sigma_j - \sigma_{j-1}|
 \end{aligned}$$

$$\downarrow \text{ If } \quad c_j + D_j \leq 1 \quad \forall j$$

$$D_j \geq 0 \quad \forall j$$

$$\downarrow c_j \geq 0 \quad \forall j$$

Then

$$\sum_{j=-\infty}^{+\infty} |\sigma_{j+1}^{n+1} - \sigma_j^{n+1}| \leq \sum_{j=-\infty}^{+\infty} (1 - q - D_j) |\sigma_{j+1} - \sigma_j|$$

$$+ \sum_{j=-\infty}^{+\infty} D_{j+1} |\sigma_{j+2} - \sigma_{j+1}| + \sum_{j=-\infty}^{+\infty} E_{j-1} |\sigma_j - \sigma_{j-1}|$$

$$\sum_{j=-\infty}^{+\infty} |\sigma_{j+1}^{n+1} - \sigma_j^{n+1}| \leq \sum_{j=-\infty}^{+\infty} (1 - q - D_j) |\sigma_{j+1} - \sigma_j| + \sum_{j=-\infty}^{+\infty} D_j |\sigma_{j+1} - \sigma_j|$$

$$+ \sum_{j=-\infty}^{+\infty} E_j |\sigma_{j+1} - \sigma_j|$$

$$= \sum_{j=-\infty}^{+\infty} |\sigma_{j+1} - \sigma_j|$$

Ex 16.2 see notes above

$$0 \leq \left\{ 1 + \frac{1}{2}(1-\nu) \left[ \frac{\phi(\theta_j)}{\theta_j} - \phi(\theta_{j-1}) \right] \right\} \leq 1$$

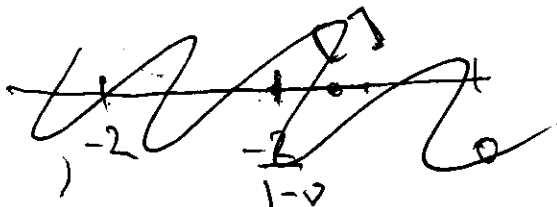
$$|\nu| \leq 1$$

$$-1 < \nu < 0$$

$$\Rightarrow 0 \geq \left\{ 1 + \frac{1}{2}(1-\nu) [ \quad ] \right\} \geq \frac{1}{\nu} = \frac{-1}{|\nu|}$$

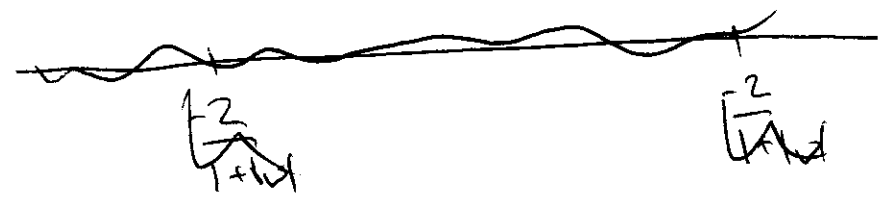
~~then,~~

$$\frac{-2}{1-\nu} [ \quad ] \geq \frac{2(1-\frac{1}{|\nu|})}{1-\nu} = \frac{2(-\nu-1)}{-\nu(1-\nu)}$$



$$\frac{2(|v|-1)}{|v|(1+|v|)} \leq [ \quad ] \leq \frac{-2}{1+|v|} \quad \text{w/ } |v| < 1$$

$$\Rightarrow \frac{-2(1-|v|)}{|v|(1+|v|)} \leq [ \quad ] \leq \frac{-2}{1+|v|}$$



$$-2 \leq \frac{-2(\frac{1}{1+v} - 1)}{1+|v|} \leq [ \quad ] \leq \frac{-2}{1+|v|} \leq 0$$

to show need that  $0 < \frac{1}{1+v} - 1 < 1$  ?  $\Leftrightarrow 1 - |v| < |v| + |v|^2$   
 $\Leftrightarrow |v|^2 + 2|v| - 1 > 0$  ?

$$(|v|-1)^2 > 0 \quad \checkmark \quad \text{true.}$$

IF  $0 < v < 1$        $0 \leq g_{-1} \leq 1$

$$\Rightarrow \quad \cancel{\frac{2(0-1)}{1-v}} \leq [ \quad ] \leq \cancel{\frac{2(v-1)}{1-v}}$$

$$\quad \frac{-2}{1-v} \leq [ \quad ] \leq 2$$

or  $\frac{2}{1-v} \geq - [ \quad ] \geq -\frac{2(v-1)}{1-v}$

?  
~~MAX~~<sup>2</sup>  
~~MIN~~<sup>2</sup>  
~~(v-1)~~<sup>2</sup>

know  $\frac{1-v-1}{1-v} < 1$

$$\Rightarrow \frac{1-v-1}{1-v} > 1$$

see 17.11 part

$$1 > \nu > 0$$

$$\frac{-2}{1-\nu} = \frac{2(0-1)}{(1-\nu)} \leq \frac{\phi(\theta_j)}{\theta_j} - \phi(\theta_{j-1}) \leq \frac{2(\nu-1)}{(1-\nu)} = \frac{2}{\nu}$$

$$1 > 1-\nu > 0$$

Then

$$\frac{-2}{\nu} \leq -\left(\frac{\phi(\theta_j)}{\theta_j} - \phi(\theta_{j-1})\right) \leq \frac{2}{1-\nu}$$

Pg 183 LeVeque

$$\frac{1}{h} \int_{x_{j-1/2}}^{x_{j+1/2}} \hat{U}_3^u(x, t_n) dx = U_j^u + B_j^u \left( \frac{x^2}{2} - x_j x \right) \Big|_{x_{j-1/2} = x_j - \frac{h}{2}}^{x_{j+1/2}}$$

$$= U_j^u + B_j^u \left[ \frac{(x_{j+1/2})^2 - (x_{j-1/2})^2}{2} - x_j (x_{j+1/2}) + x_j (x_{j-1/2}) \right]$$

||

0 ✓.

$$U_j^{n+1} = U_j^n - \frac{k}{h} [F(U_j^n, U_{j+1}^n) - F(U_{j-1}^n, U_j^n)]$$

$$\text{w/ } F(U_j^n, U_{j+1}^n) = \frac{1}{k} \int_{x_j}^{x_{j+1}} f(\tilde{U}^n(x, t)) dx$$

$$\text{let } \tilde{U}^n(x, t) = U_j^n + \delta_j^n (x - x_j)$$

$$\text{If } \delta_j^n = \frac{U_{j+1}^n - U_j^n}{h}$$

$$\text{Then if } f(u) = au \quad (a > 0)$$

$$F(U_j^n, U_{j+1}^n) = \frac{a}{k} \left[ h U_j^n + \delta_j^n \left( x_j + \frac{h}{2} - x_j \right) \right]$$



Pg 185 L/Equ

$$\nu = \frac{ka}{h}$$

$$F(U_{ij}) = a U_j + \frac{1}{2} a (1 - \nu) h B_j$$

Pg 186 L/Equ

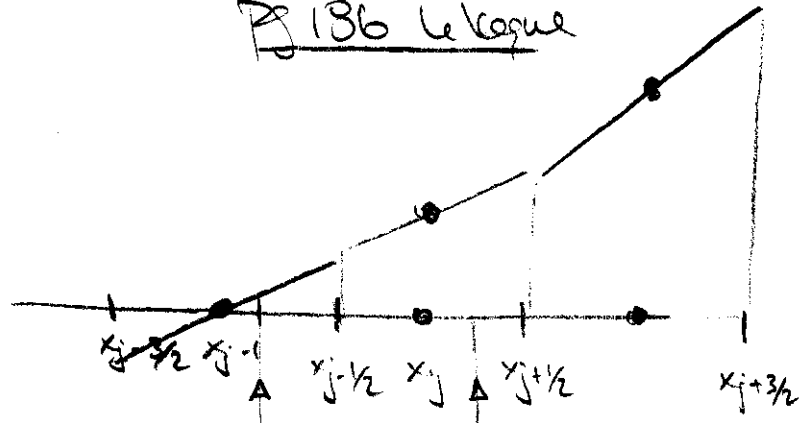
$$F(U_{ij}) = a U_{j_1} + \frac{1}{2} a (\operatorname{sgn}(\nu) - \nu) (h B_{j_1})$$

)

)

)

Ex 16.3



By step 3

$$\sigma_j^{n+1} = \int_{x_{j-1/2}}^{x_{j+1/2}} \tilde{u}(x, t_{n+1}) dx = \int_{x_{j-1/2}}^{x_{j+1/2}} \tilde{u}^n(x - ak, t_n) dx$$

$$= \int_{x_{j+1/2} - ak}^{x_{j+1/2} - ah} \tilde{u}^n(x, t_n) dx = \int_{x_{j-1/2} - ah}^{x_{j+1/2} - ah} \tilde{u}^n(x, t_n) dx$$

If  $a > 0$

since  $ak = ah$

$$= \int_{x_{j-1/2} - ah}^{x_{j-1/2}} (\sigma_{j-1}^n + \beta_{j-1}^n (x - x_{j-1})) dx$$

$d > 1$   
so  $ah < h$

$$+ \int_{x_{j-1/2}}^{x_{j+1/2} - ah} (\sigma_j^n + \beta_j^n (x - x_j)) dx$$

$$= \sigma_{j-1}^n (x_{j-1/2} - x_{j-1/2} + ah) + \beta_{j-1}^n \int_{x_{j-1/2} - ah}^{x_{j-1/2}} (x - x_{j-1}) dx$$

$$+ T_j^n (x_{j+1/2} - \nu h - x_{j-1/2}) + B_j^n \int_{x_{j-1/2}}^{x_{j+1/2} - \nu h} (x - x_j) dx$$

$$= T_{j-1}^n \nu h + B_{j-1}^n \int_{x_{j-1/2} - \nu h - x_{j-1}}^{x_{j-1/2} - x_{j-1}} v dv + T_j^n (h - \nu h) + B_j^n \int_{x_{j-1/2} - x_j}^{x_{j+1/2} - \nu h - x_j} v dv$$

$$= T_{j-1}^n \nu h + B_{j-1}^n \int_{\nu/2 - \nu h}^{\nu/2} v dv + T_j^n h(1 - \nu) + B_j^n \int_{-\nu/2}^{\nu/2 - \nu h} v dv$$

$$= T_{j-1}^n \nu h + B_{j-1}^n \left( \frac{h^2}{24} - \frac{1}{2} \left( \frac{h}{2} - \nu h \right)^2 \right)$$

$$+ T_j^n h(1 - \nu) + B_j^n \left( \frac{1}{2} \left( \frac{h}{2} - \nu h \right)^2 - \frac{h^2}{24} \right)$$

$$= \nu h T_{j-1}^n + B_{j-1}^n h^2 \left[ \frac{1}{8} - \frac{1}{8} (1 - 2\nu)^2 \right]$$

$$+ T_j^n h(1 - \nu) + \frac{B_j^n}{8} \left[ (1 - 2\nu)^2 - 1 \right] h^2$$

$$= \nu h U_{j-1}^n + \frac{h^2 \nu}{\delta_{j-1}} \left[ X - 1 + 4\nu - 4\nu^2 \right]$$

$$+ h(1-\nu) U_j^n + \frac{h^2 \nu}{\delta_j} \left[ X - 4\nu + 4\nu^2 - X \right]$$

$$= \nu h U_{j-1}^n + \frac{h^2 \nu}{2} \delta_{j-1} (1-\nu) + h U_j^n - h\nu U_j^n$$

$$- \frac{h^2 \nu}{2} (1-\nu) \delta_j^n$$

Since Average is defined as  $\frac{1}{h} \int_{x-\frac{h}{2}}^{x+\frac{h}{2}} \bar{U}(x, t_{n+1}) dx$  + I forgot the  $\frac{1}{h}$

we get

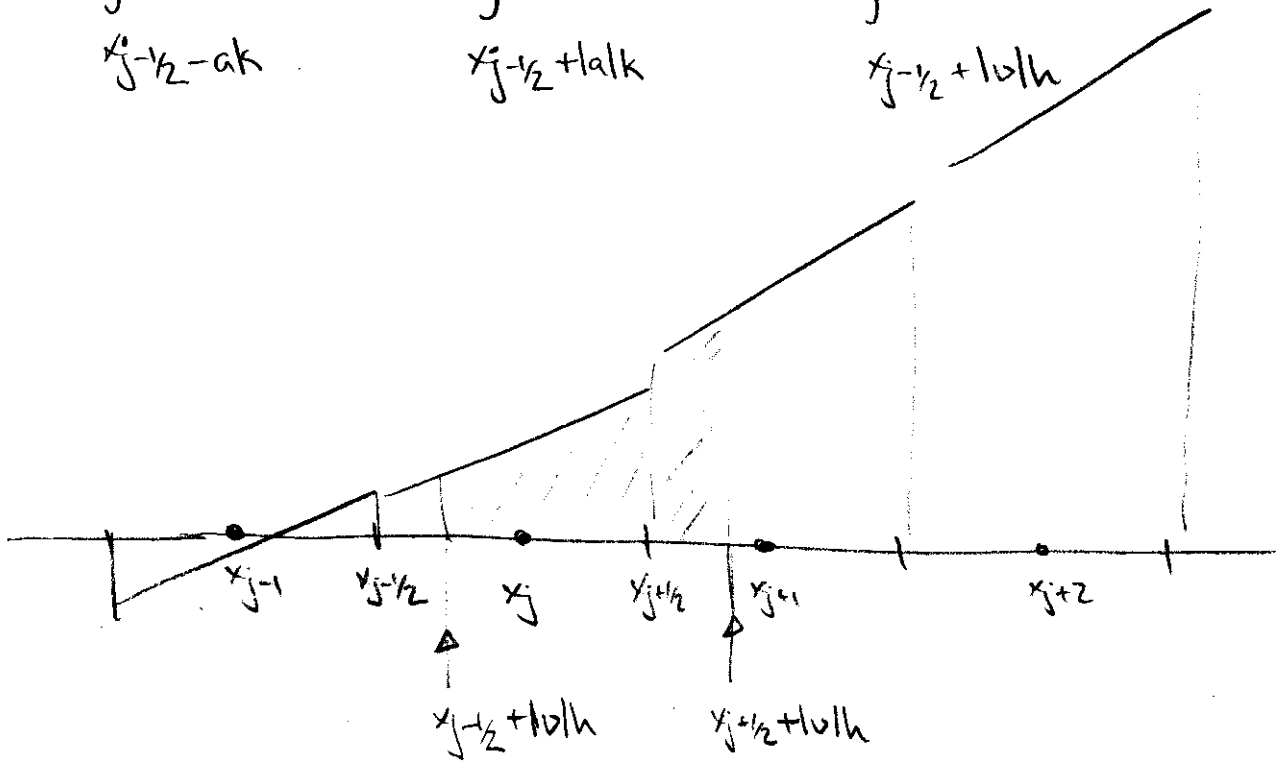
$$U_j^{n+1} = U_j^n + \nu(U_{j-1}^n - U_j^n) + \frac{h\nu}{2}(1-\nu)(\delta_{j-1}^n - \delta_j^n)$$

or

$$U_j^{n+1} = U_j^n - \nu(U_j^n - U_{j-1}^n) - \frac{1}{2}\nu(1-\nu)(h\delta_j^n - h\delta_{j-1}^n) \quad \text{eq 16.45}$$

If  $a < 0$  Then

$$U_j^{n+1} = \int_{x_{j-1/2} - ak}^{x_{j+1/2} - ak} \tilde{U}^n(x, t_n) dx = \int_{x_{j-1/2} + l|a|h}^{x_{j+1/2} + l|a|h} \tilde{U}^n(x, t_n) dx = \int_{x_{j-1/2} + l|a|h}^{x_{j+1/2} + l|a|h} \tilde{U}^n(x, t_n) dx$$



$$= \int_{x_{j-1/2} + l|a|h}^{x_{j+1/2}} (U_j^n + \beta_j^n (x - x_j)) dx + \int_{x_{j+1/2}}^{x_{j+1/2} + l|a|h} (U_{j+1}^n + \beta_{j+1}^n (x - x_{j+1})) dx$$

$$= U_j^n (x_{j+1/2} - x_{j-1/2} - l|a|h) + \beta_j^n \int_{x_{j-1/2} + l|a|h}^{x_{j+1/2}} (x - x_j) dx$$

$$+ U_{j+1}^n (x_{j+1/2} + l|a|h - x_{j+1/2}) + \beta_{j+1}^n \int_{x_{j+1/2}}^{x_{j+1/2} + l|a|h} (x - x_{j+1}) dx$$

$$= U_j^n (h - |v| h) + \delta_j^n \int_{-\frac{h}{2} + |v|h}^{\frac{h}{2}} v dv + U_{j+1}^n h |v| + \delta_{j+1}^n \int_{\frac{h}{2}}^{-\frac{h}{2} + |v|h} v dv$$

$$= h U_j^n (1 - |v|) + \frac{\delta_j^n}{2} \left( \frac{h^2}{4} - \left( -\frac{h}{2} + |v|h \right)^2 \right) + U_{j+1}^n h |v| + \frac{\delta_{j+1}^n}{2} \left[ \left( -\frac{h}{2} + |v|h \right)^2 - \frac{h^2}{4} \right]$$

$$= h U_j^n (1 - |v|) + \frac{\delta_j^n}{2} h^2 \left[ \frac{1}{4} - \left( \frac{1}{4} - |v| + |v|^2 \right) \right] + U_{j+1}^n h |v|$$

$$+ \frac{\delta_{j+1}^n}{2} h^2 \left[ \left( \frac{1}{4} - |v| + |v|^2 \right) - \frac{1}{4} \right]$$

$$= h U_j^n (1 - |v|) + \frac{\delta_j^n}{2} h^2 |v| (1 - |v|) + U_{j+1}^n h |v| - \frac{\delta_{j+1}^n}{2} h^2 |v| (1 - |v|)$$

÷ by h give

$$U_j^{n+1} = U_j^n (1 - |v|) + U_{j+1}^n |v| + \frac{h}{2} \delta_j^n |v| (1 - |v|) - \frac{h}{2} \delta_{j+1}^n |v| (1 - |v|)$$

$$= U_j^n + |v| (U_{j+1}^n - U_j^n) + \frac{1}{2} |v| (1 - |v|) (h \delta_j^n - h \delta_{j+1}^n)$$

$$\Rightarrow U_j^{n+1} = U_j^n + |v| (U_{j+1}^n - U_j^n) - \frac{|v|(1 - |v|)}{2} (h \delta_{j+1}^n - h \delta_j^n)$$

Now: when  $a > 0$ ,  $v > 0$

$$U_j^{n+1} = U_j^n - v(U_j^n - U_{j-1}^n) - \frac{1}{2}v(1-v)(h\delta_j^n - h\delta_{j-1}^n)$$

↓ when  $a < 0$ ,  $v < 0$

$$U_j^{n+1} = U_j^n + |v|(U_{j+1}^n - U_j^n) - \frac{|v|}{2}(1-|v|)(h\delta_{j+1}^n - h\delta_j^n)$$

These can be combined into 1 formula as

$$U_j^{n+1} = U_j^n - v(U_{j_i}^n - U_{j_{i-1}}^n) - \frac{|v|}{2}(\text{sign}(v) - v)(h\delta_{j_i}^n - h\delta_{j_{i-1}}^n)$$

$$j_i = \begin{cases} j & a > 0 \\ j+1 & a < 0 \end{cases}$$

I got  $|v|$  rather than  $v$  in the above

Method 16.48 is  $\sigma_j^{n+1} = \sigma_j^n - \nu(\sigma_j^n - \sigma_{j-1}^n) - \frac{1}{2}\nu(\text{sgn}(\nu) - \nu)(h\beta_{j1} - h\beta_{j-1})$

used to update the  $p$ th entry:

$$V_{pj}^{n+1} = V_{pj}^n - \nu_p(V_{p,jp} - V_{p,jp-1}) - \frac{1}{2}\nu_p(\text{sgn}(\nu_p) - \nu_p)(h\beta_{p,jp} - h\beta_{p,jp-1})$$

$$\beta_{pj} = \frac{1}{h} \min \text{mod}(V_{p,j+1} - V_{pj}, V_{pj} - V_{p,j-1}) = \frac{1}{h} \min \text{mod}(\alpha_{pj}, \alpha_{p,j-1})$$

Multiply by  $r_p$  & sum over  $p$ .

$$\sigma_j^{n+1} = \sigma_j^n - \frac{k}{h} \left[ \sum_{p=1}^m (V_{p,jp} r_p - V_{p,jp-1} r_p) + \frac{1}{2} \sum_{p=1}^m \nu_p (\text{sgn}(\nu_p) - \nu_p) (h\beta_{p,jp} r_p - h\beta_{p,jp-1} r_p) \right]$$

Defining  $\beta_{p,jp} r_p = \beta_{pj}$  ↓ same

$$F_L(\sigma; j) = \sum_{p=1}^m V_{p,jp} r_p$$

$$\sigma_j^{n+1} = \sigma_j^n - \frac{k}{h} \left[ \sum_{p=1}^m V_{p,jp} r_p + \frac{1}{2} \sum_{p=1}^m \nu_p (\text{sgn}(\nu_p) - \nu_p) h\beta_{pj} - \sum_{p=1}^m V_{p,jp-1} r_p - \frac{1}{2} \sum_{p=1}^m \nu_p (\text{sgn}(\nu_p) - \nu_p) h\beta_{p,jp-1} \right]$$



Thus flux  $F(\sigma; j) = F_L(\sigma; j)$

$$+ \frac{1}{2} \sum_{p=1}^m \Delta_p (\text{sgn}(v_p) - v_p) h \beta_{pj}$$

Flux 16.40 is

$$F(\sigma; j) = F(\sigma; j) + \frac{1}{2} \sum_{p=1}^m \phi(\theta_{pj}) (\text{sgn}(v_p) - v_p) \Delta_p \alpha_{pj} r_p$$

∴ the sum flux is obtained if we choose  $\beta_{pj} = \frac{\phi(\theta_{pj}) \alpha_{pj} r_p}{h}$

eq 16.61

### Ex 16.4

By 16.58 + 16.56

$$\beta_{pj} = \beta_{pj} r_p = \frac{1}{h} \text{minmod}(\alpha_{pj}, \alpha_{pj-1}) r_p = \text{LHS.}$$

By 16.37

∴  $\phi$  minmod limiter 16.53

$$\theta_{pj} = \frac{\alpha_{pj'}}{\alpha_{pj}}$$

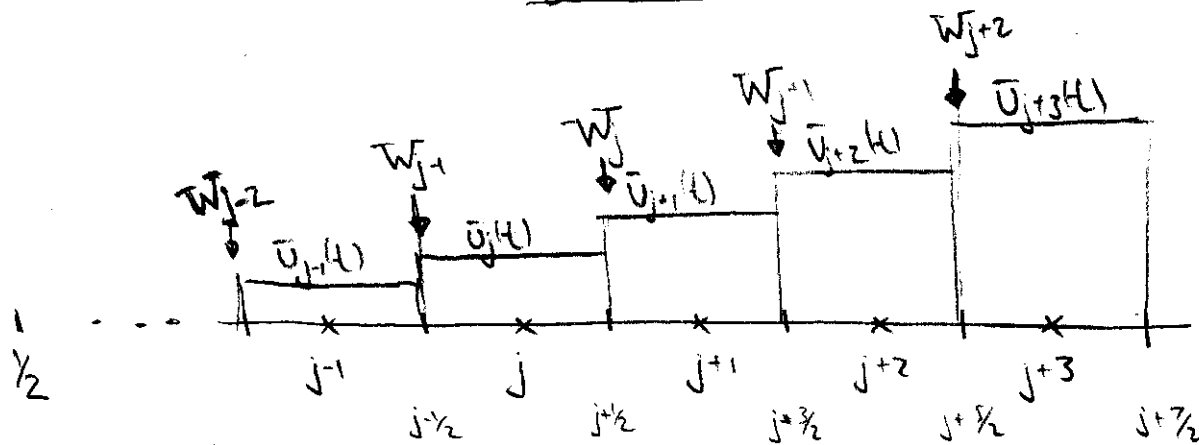
w/  $j' = j - \text{sgn}(v_p)$

Then R.H.S is

$$\frac{1}{h} \phi_{\text{minmod}}\left(\frac{\alpha_{pj'}}{\alpha_{pj}}\right) \alpha_{pj} r_p$$

$$\phi\left(\frac{\alpha_{P_i'}}{\alpha_{P_j'}}\right) = \begin{cases} 0 & \text{if } \frac{\alpha_{P_i'}}{\alpha_{P_j'}} < 0 \\ \frac{\alpha_{P_i'}}{\alpha_{P_j'}} & \text{if } 0 \leq \frac{\alpha_{P_i'}}{\alpha_{P_j'}} \leq 1 \quad \alpha_{P_i'} \leq \alpha_{P_j'} \\ 1 & \text{if } \frac{\alpha_{P_i'}}{\alpha_{P_j'}} \geq 1 \quad \alpha_{P_i'} \geq \alpha_{P_j'} \end{cases}$$

This doesn't seem to be  $\min \text{mod}(\alpha_{P_i'}, \alpha_{P_j'})$  as I think it should.



Given  $\{ \bar{u}_j(t) : j \in (-\infty, +\infty) \}$  how can I reconstruct accurately

values of  $u$  to high order?

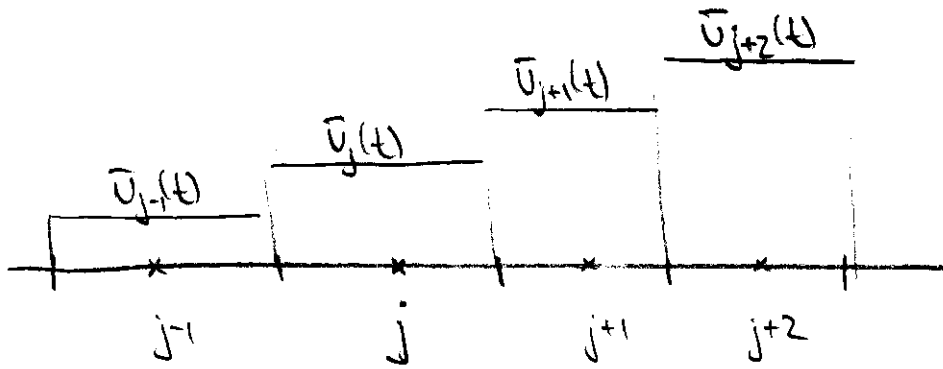
Define an auxiliary fn  $w(x)$  as  $w(x) = \int_{x_{j-1/2}}^x u(\xi, t) d\xi$

Note then  $w'(x) = u(x, t)$ . Thus if one can approximate  $w(x)$  to high order one can differentiate this fn & obtain an accurate approximation to  $u(x, t)$ .

define  $W_j = w(x_{j+1/2}) = \int_{x_{j-1/2}}^{x_{j+1/2}} u(\xi, t) d\xi = h \sum_{i=1}^j \bar{u}_i(t)$  &  $W_j$  is exact at the  $1/2$  nodes if  $\bar{u}_i(t)$  is known exactly.

Now it is obvious that we have the exact values for  $w(x)$  at the  $1/2$  nodes.

Ex 17.1



$q=2 \Rightarrow$  We wish to approximate  $u$  on  $[x_{j-1/2}, x_{j+1/2}]$  based on a derivative of an interpolating polynomial for  $w(x) = \int_{x_{j-1/2}}^x u(\xi, t) d\xi$

The ENO scheme requires that we choose the minimum between the divided differences of

$$\{w_{j-2}, w_{j-1}, w_j\} + \{w_{j-1}, w_j, w_{j+1}\}$$

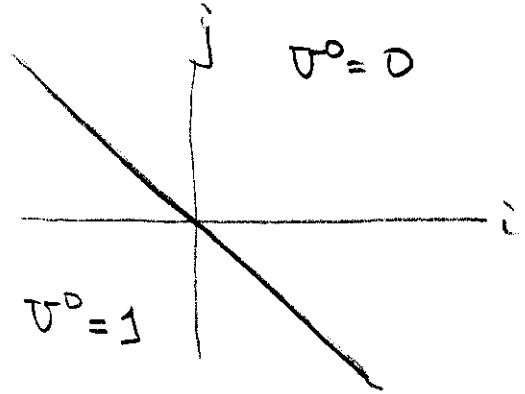
$$\begin{array}{l} w_{j-2} \\ w_{j-1} \\ w_j \end{array} \left\{ \begin{array}{l} \frac{w_{j-2} - w_{j-1}}{h} \\ \frac{w_{j-1} - w_j}{h} \end{array} \right\} \left\{ \begin{array}{l} \frac{w_{j-2} - w_{j-1} - w_{j-1} + w_j}{h^2} \\ \frac{w_{j-2} - 2w_{j-1} + w_j}{h^2} \end{array} \right.$$

performing the same  $\div$  difference we obtain

Ex 18.1

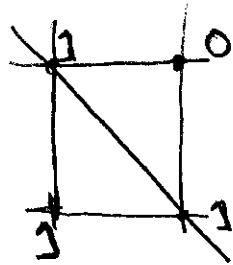
$$u_t + u_x + u_y = 0$$

$$U_{ij}^0 = \begin{cases} 1 & i+j < 0 \\ 0 & i+j \geq 0 \end{cases}$$



Solving 1st  $u_t + f(u)_x = 0$  + then

$$u_t + g(u)_y = 0$$



Not sure how to do?

$$e^{-tB\partial_y} e^{-tA\partial_x} = (1 - tB\partial_y + \frac{t^2 B^2 \partial_y^2}{2} - \dots) (1 - tA\partial_x + \frac{t^2 A^2 \partial_x^2}{2} - \dots)$$

$$= 1 - t(A\partial_x + B\partial_y) + \frac{t^2}{2} (A^2 \partial_x^2 + B^2 \partial_y^2 + 2BA\partial_y \partial_x) + O(t^3)$$

Now

$$(A\partial_x + B\partial_y)^2 = (A\partial_x + B\partial_y)(A\partial_x + B\partial_y)$$

$$= A^2 \partial_x^2 + AB\partial_x \partial_y + BA\partial_y \partial_x + B^2 \partial_y^2$$

Thus we shall add  $AB\partial_x \partial_y$  to attempt to get this second term

$$\Rightarrow e^{-tB\partial_y} e^{-tA\partial_x} = 1 - t(A\partial_x + B\partial_y) + \frac{t^2}{2} (A^2 \partial_x^2 + AB\partial_x \partial_y + BA\partial_y \partial_x + B^2 \partial_y^2 - AB\partial_x \partial_y + BA\partial_y \partial_x)$$

$$+ O(t^3)$$

$$= 1 - t(A\partial_x + B\partial_y) + \frac{t^2}{2} [(A\partial_x + B\partial_y)^2 - (AB - BA)\partial_x \partial_y]$$

$$+ O(t^3)$$

$$= 1 - t(A\partial_x + B\partial_y) + \frac{t^2}{2} (A\partial_x + B\partial_y)^2 - \frac{t^2}{2} (AB - BA)\partial_x \partial_y + O(t^3)$$

$$= e^{-k(A\partial_x + B\partial_y)} - \frac{k^2}{2}(AB - BA)\partial_x\partial_y + O(k^3)$$

Because I can add all the higher order terms I need w/ the  $O(k^3)$

$$e^{-\frac{1}{2}kA\partial_x} e^{-kB\partial_y} e^{-\frac{1}{2}kA\partial_x} =$$

$$\left(1 - \frac{kA\partial_x}{2} + \frac{(\frac{kA\partial_x}{2})^2}{2} - \frac{1}{3!}(\frac{kA\partial_x}{2})^3 + \dots\right) \cdot$$

$$\left(1 - kB\partial_y + \frac{1}{2}(kB\partial_y)^2 - \frac{1}{3!}(kB\partial_y)^3 + \dots\right) \cdot$$

$$\left(1 - \frac{1}{2}kA\partial_x + \frac{1}{2}(\dots)\right)$$

$$= 1 - \frac{k}{2}A\partial_x - kB\partial_y - \frac{1}{2}kA\partial_x + 2\frac{1}{2}k$$

$$(A\partial_x + B\partial_y)^2 = A^2\partial_x^2 + AB\partial_x\partial_y + BA\partial_y\partial_x + B^2\partial_y^2$$

$$\sigma^n = H_{F/2}^x H_F^y H_{F/2}^x \sigma^{n-1} = H_{F/2}^x (H_F^y H_{F/2}^x) (H_{F/2}^x H_F^y H_{F/2}^x) \sigma^{n-2}$$

$$= H_{F/2}^x (H_F^y H_{F/2}^x) H_F^y H_{F/2}^x \sigma^{n-2}$$

$$= H_{F/2}^x (H_F^y H_{F/2}^x) (H_F^y H_{F/2}^x H_F^y H_{F/2}^x) \sigma^{n-3}$$

$$= (H_F^y H_{F/2}^x)^2 H_F^y H_{F/2}^x \sigma^{n-3}$$

...

$$\sigma^n = H_{F/2}^x (H_F^y H_{F/2}^x)^{n-1} H_F^y H_{F/2}^x \sigma^0$$



Ex 18.2

$$U^{n+1} = H_{1/2}^x H_k^y H_{1/2}^x U^n$$

We are told that  $H_k^x = H_f^x + O(k^2)$  w/  $H_f^x$  the exact solution operator.

Thus

$$U^{n+1} = \left( H_{1/2}^x + O(k^2/4) \right) \left( H_k^y + O(k^2) \right) \left( H_{1/2}^x + O(k^2/4) \right)$$

$$= H_f^x H_k^y H_{1/2}^x + H_{1/2}^x$$