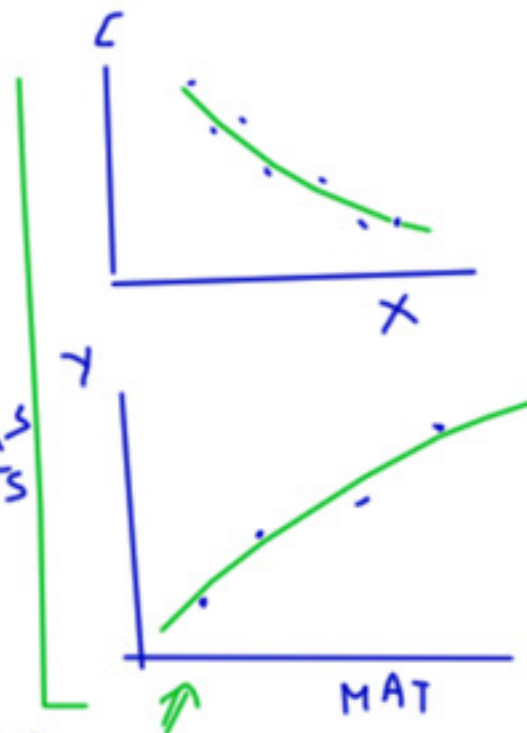


1 BONDS INTRO

- US: DEFAULT FREE, CASHFLOW KNOWN $P_t = E_t(M_{t+1} \cdot 1)$

• DATA PLOT

- UP + DOWN WITH INFLATION
- WITH RECESSIONS!
- SPREAD HIGH IN RECESSION, LOW IN BOOMS
- "LEVEL", "SLOPE" CAPTURE MOST MOVEMENTS
- "YIELD CURVE" / "FORWARD CURVE"
- CAN SLOPE UP/DOWN
- WHY? (ECONOMICS) - MODEL JUST LIKE BLACK-SCHOLES?
 - PRICE ALL MATURITIES
 - PRICE / HEDGE TS OPTIONS



BONDS 2. DEFINITIONS

ZERO-COUPON BOND → PAYS \$1 AT $t+n$

OBJECT	IDEA	EXAMPLE
PRICE Maturity ↙ DATE		0.9 (\$90 REAL WORLD)
log PRICE	"10% DISCOUNT"	$-0.1 = \log(0.9)$
YIELD	"AVERAGE RETURN" "BOND DISCOUNT RATE"	$P_t^{(2)} = -0.1$ "10% DISCOUNT" $y_t^{(2)} = +0.05$ "5%/YEAR"
FORWARD	"RATE YOU CAN LOCK IN TODAY, TO BORROW AT n , PAY BACK $n+1$ "	$P_t^{(2)} = -0.1$ $P_t^{(1)} = -0.07 \rightarrow f_t^{(2)} = 0.05$
RETURN		
EXCESS RETURN	WARNING: $\log(R - R^f) \neq \log(R) - \log(R^f)$	

③ COUPON BONDS $P_t = P_t^{(1)} \cdot C + P_t^{(2)} \cdot C + \dots + P_t^{(n)} \cdot [C+1]$ ← ZEROS LIKE CONTINGENT CLAIMS

BONDS 3 EXPECTATIONS HYPOTHESIS

- WHY DOES YIELD/FORWARD CURVE SLOPE?
- IF $y_t^{(1)} = 0.1\%$, $y_t^{(2)} = 2\%$. IS $y_t^{(1)}$ A BETTER DEAL?
- EH ≈ "RISK NEUTRAL", "RANDOM WALK". NO. $y_t^{(1)}$, $y_t^{(2)}$
GIVE THE SAME RETURN IN YEAR 1, RATES WILL RISE.



$r_0 \rightarrow n$
A. $-P_t^{(n)} - E_t(y_t^{(1)} + y_{t+1}^{(1)} + \dots + y_{t+n-1}^{(1)} - y_t^{(n)}) \Rightarrow y_t^{(n)} = E_t \left(\frac{1}{n} (y_t^{(1)} + y_{t+1}^{(1)} + \dots + y_{t+n-1}^{(1)}) \right)$ [+RISK PREM.]

B. $y_t^{(1)} = E_t(r_{t+1}^{(1)})$; $E_t(r_{t+1}^{(1)}) = 0$ [+RISK PREMIUM]

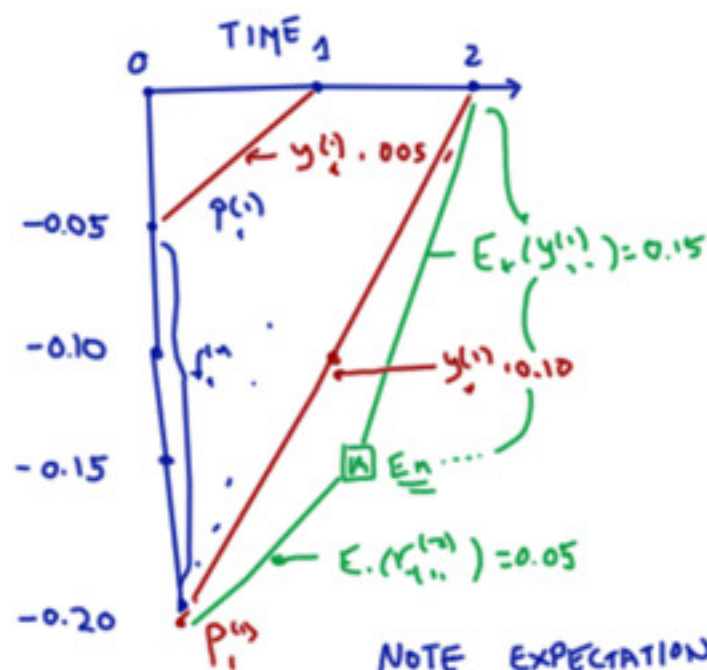
C. $f_t^{(n)} = E_t(y_{t+n-1}^{(1)})$

FACT
A ↔ B ↔ C

3.1 EXPECTATIONS EXAMPLE

$$\begin{aligned}
 P_t^{(1)} = -0.05 &\Rightarrow y_t^{(1)} = 0.05 \\
 P_t^{(2)} = -0.20 &\Rightarrow y_t^{(2)} = 0.10 \\
 &\Rightarrow f_t^{(2)} = 0.15
 \end{aligned}$$

$$\begin{aligned}
 E_t(y_{t+1}^{(1)}) &= 0.15 \\
 E_t(P_{t+1}^{(1)}) &= -0.15 \\
 E_t(r_{t+1}^{(2)}) &= 0.05
 \end{aligned}$$



$$\begin{aligned}
 & \left. \begin{aligned} & \uparrow \\ & \text{=} \\ & \text{=} \end{aligned} \right\} = \\
 & y_t^{(2)} = E_t(r_{t+1}^{(2)}) \quad \left\{ \begin{aligned} & f_t^{(2)} = E_t(y_{t+1}^{(1)}) \\ & \Leftrightarrow y_t^{(2)} = \frac{1}{2} E_t(y_{t+1}^{(1)}, y_{t+1}^{(2)}) \end{aligned} \right.
 \end{aligned}$$

ALL TRM \square WILL END UP ELSEWHERE \square

ft
 $r^{EURO} = 5\%$, $r^{US} = 1\%$ \rightarrow Buy?

EH \rightarrow EURO \downarrow 4%!

$$E(S_{t+1}) - S_t = r_t^{EU} - r_t^{US} [+R.P.]$$

BOTTOM LINE

- 1) STARTING PLACE "ALLER =" HAS SURPRISING IMPLICATIONS!
- 2) IDENTITY "Y MUST MOVE SO X DOES NOT"
 EITHER $E_t(r_{t+1}^{(2)})$ IS HIGH OR $E_t(y_{t+1}^{(1)} - y_{t+1}^{(2)})$ IS HIGH
 E_t : WHAT MUST MOVE IF ER DOES NOT MOVE

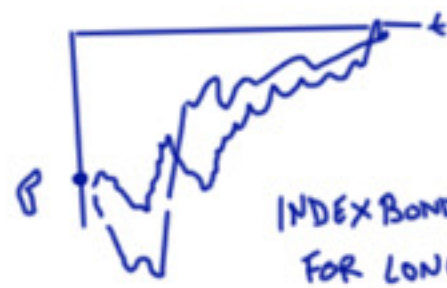
BONDS & RISK PREMIUM INTRODUCTION

• WHICH IS RISKIER?

- $r_{t+1}^{(2)}$ vs $y_t^{(1)}$

- $y_t^{(2)}$ vs $y_t^{(1)}$, $y_{t+1}^{(1)}$, ... $y_{t+n}^{(1)}$

- WHAT IF REAL $y_t^{(1)}$ ARE CONSTANT, INFLATION VARIES?



INDEX BONDS ARE RISK FREE FOR LONG HORIZON INVESTORS. IGNORE PRICE CHANGES

REAL YIELD CURVE SHOULD SLOPE DOWN



BONDS 4 CONT'D

• $E_t(R_{t+n}^A) = \gamma \text{COV}_t(R_{t+n}^R, \Delta C_{t+n})$
 $= -\gamma \eta (\text{cov}_t(y_{t+n}^{(n)}, \Delta C_{t+n}))$

ISSUE "MARKET PRICE OF INTEREST RATE RISK"

DOES A SHOCK $y_{t+n}^{(n)} \uparrow$ COME W. GOOD OR BAD ΔC_{t+n} ?

$\downarrow E(R^A) < 0!$

- EMPIRICAL EVIDENCE [NO GOOD THEORY... YET]
- "EXPECTATIONS" - STRICT - NO R.P.
 - R.P.'S SHALL, CONSTANT OVERTIME
 - NOT? R.P. VARIES OVERTIME

FX $S_{t+n} - S_t = a + b(r_t^f - r_t^d) + \epsilon_{t+n}$

SHOULD BE -1
 IS 0 - +1! "CARRY TRADE"

"UNCOVERED INTEREST PARITY" PUZZLE
 RISK PREMIUM TO HOLD HIGH r CURRENCIES, BORROW LOW

5

BONDS 5. FACTS: FAHA-BLISS

- LOOK AT DATA. GOOD!
- REGRESSIONS:

$r_{t+n}^{(n)} = a + b(f_t^{(n)} \cdot y_t^{(n)}) + \epsilon_{t+n}$

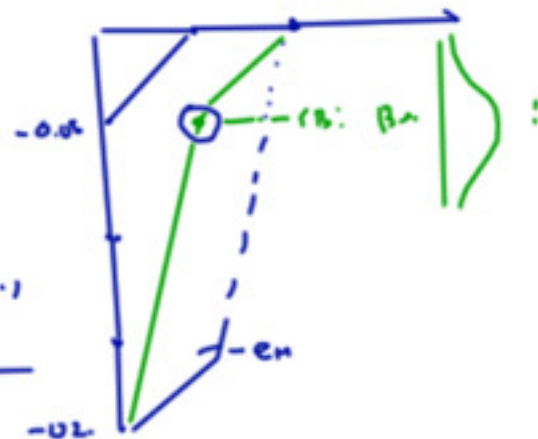
n	b	$\sigma(b)$	R ²
2	0.83	0.27	0.11
3	1.14	0.35	0.13
4	1.38	0.43	0.15
5	1.05	0.49	0.07

FORECAST 1 YEAR R ON n YEAR BONDS

$y_{t+n}^{(n)} - y_t^{(n)} = a + b(f_t^{(n)} \cdot y_t^{(n)}) + \epsilon_{t+n}$

n	b	$\sigma(b)$	R ²
1	0.17	0.27	0.01
2	0.53	0.33	0.05
3	0.89	0.26	0.14
4	0.92	0.17	0.17

FORECAST ONE YEAR RATE n-1 YEARS FROM NOW



• $f_t^{(n)} - y_t^{(n)}$ ALMOST 1-1 TO ER

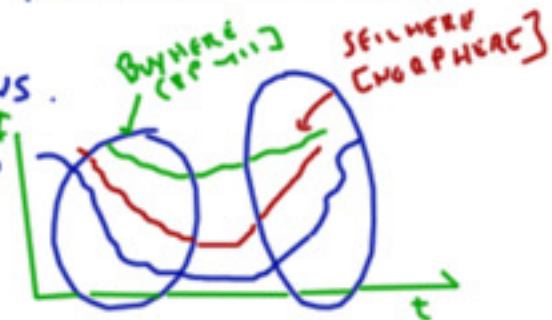
(GRAPH)

• WHY DID EYES FAIL? • YOU GET PAID TO HOLD INTEREST RATE RISK IN RECESSIONS.

• MORE RECENT (ME) $r_{t+n}^{(n)} = a + b(f_t^{(n)} \cdot y_t^{(n)}) + c \cdot Z_t + \epsilon_{t+n}$ R² UP TO 0.44. CAN SENSE THE END

• $f_t^{(2)} - y_t^{(2)}$ ALMOST 0 TO $\Delta y_t^{(2)}$ \Rightarrow EN 100% WRONG! AT 1-YEAR

• $f_t^{(5)} - y_t^{(5)}$ ALMOST 1 TO $\Delta y_{t+4}^{(5)}$ \Rightarrow EN 100% RIGHT! AT 5-YEAR



BONDS 6 STATISTICAL FACTOR MODELS

• IDEA: LIKE FF3F

$$y_t = \begin{bmatrix} y_t^{(1)} \\ y_t^{(2)} \\ y_t^{(3)} \\ \vdots \\ y_t^{(n)} \end{bmatrix} = \begin{bmatrix} | \\ | \\ | \\ | \\ | \end{bmatrix} \text{ "level", } \begin{bmatrix} -2 \\ -1 \\ 0 \\ 1 \\ 2 \end{bmatrix} \text{ "slope" + SMALL } \epsilon_t \Rightarrow \text{HIGH } R^2$$

LOADINGS
FACTORS

LEVEL = $\frac{1}{n} \sum y_t^{(n)}$ SLOPE = $y_t^{(5)} \cdot y_t^{(1)} - (y_t^{(1)} \cdot y_t^{(5)})$

• PRODUCE? CHOOSE WEIGHTS (LOADINGS) TO MAX R^2

? OLS

$$\Sigma = \text{cov}(y, y) = Q \Lambda Q' = \begin{bmatrix} | & | & | \\ q_1 & q_2 & q_r \\ | & | & | \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_r \end{bmatrix} \begin{bmatrix} \cdot & q_1 & - \\ - & q_2 & - \\ - & q_r & - \end{bmatrix}$$

6

FACTORS

$$Q Q' = Q' Q = I$$

$$x_t = Q' y_t \Rightarrow \text{cov}(x, x) = Q' \Sigma Q = \Lambda$$

FACTOR MODEL

$$y_t = Q x_t = \begin{bmatrix} | & | & | \\ q_1 & q_2 & q_5 \\ | & | & | \end{bmatrix} x_t = \begin{bmatrix} | \\ | \\ | \end{bmatrix} x_t^1 + \begin{bmatrix} | \\ | \\ | \end{bmatrix} x_t^2 + \dots + \begin{bmatrix} | \\ | \\ | \end{bmatrix} x_t^5$$

$[q_i]$ ARE WEIGHTS $x_t^i = [-q_i -] \begin{bmatrix} y_t \\ 1 \end{bmatrix}$

FACTORS ARE UNCORRELATED

$[q_i]$ ARE LOADINGS

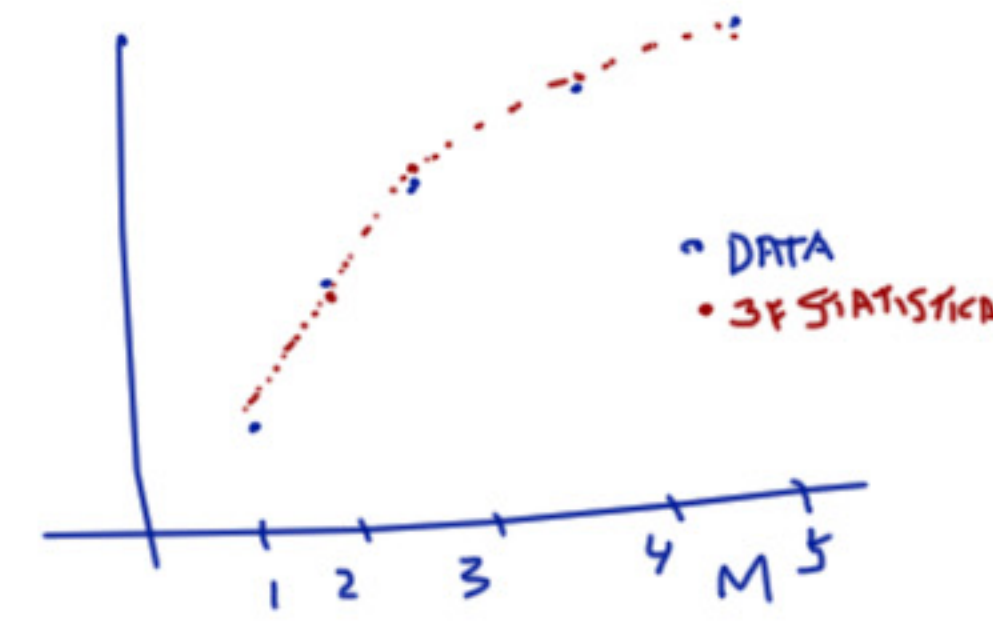
$$y_t = \begin{bmatrix} q_1 \end{bmatrix} x_t^1 + \begin{bmatrix} q_2 \end{bmatrix} x_t^2 + \underbrace{\begin{bmatrix} q_3 \end{bmatrix} x_t^3 + \dots + \begin{bmatrix} q_r \end{bmatrix} x_t^r}_{\Sigma_t}$$

↑ biggest ϵ^2

- LEAVE OUT SOME FACTORS?
 - LOADINGS ARE OLS
 - EACH ϵ MAX R^2 !

6.1 STATISTICAL FACTOR MODEL IN DATA

- BONDS
- FF25



7 TERM STRUCTURE MODEL WITH EXPECTATIONS HYPOTHESIS

- SHOWS IDEA
- "SHORT RATE PROCESS"
- NEED TO GO BEYOND INCLUDED BONDS
- "ARBITRAGE"

$$y_{t+n}^{(1)} - \delta = \phi(y_t^{(1)} - \delta) + \varepsilon_{t+1}$$

- FIND OTHERS BY EM

$$f_t^{(2)} = E_t y_{t+1}^{(1)} = \delta + \phi(y_t^{(1)} - \delta)$$

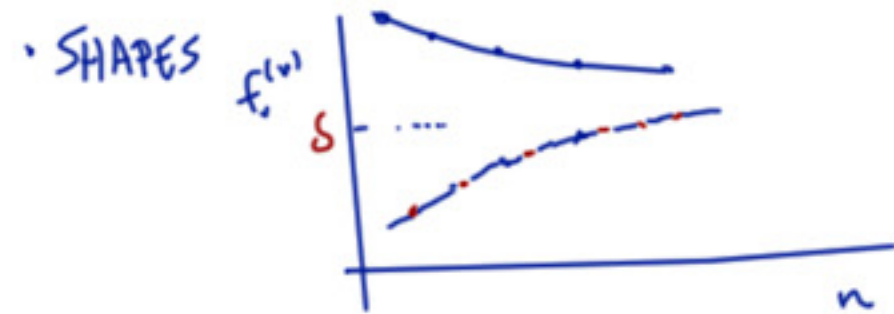
$$f_t^{(3)} = E_t(y_{t+2}^{(1)}) = \delta + \phi^2(y_t^{(1)} - \delta)$$

$$f_t^{(4)} = E_t(y_{t+m}^{(1)}) = \delta + \phi^3(y_t^{(1)} - \delta)$$

THEN

$$y_t^{(n)} = \frac{1}{n} (y_t^{(1)} + f_t^{(2)} + \dots + f_t^{(n)}) = \delta + (1 + \phi + \phi^2 + \dots + \phi^{n-1})(y_t^{(1)} - \delta)$$

- AN EXACT 1 FACTOR MODEL!

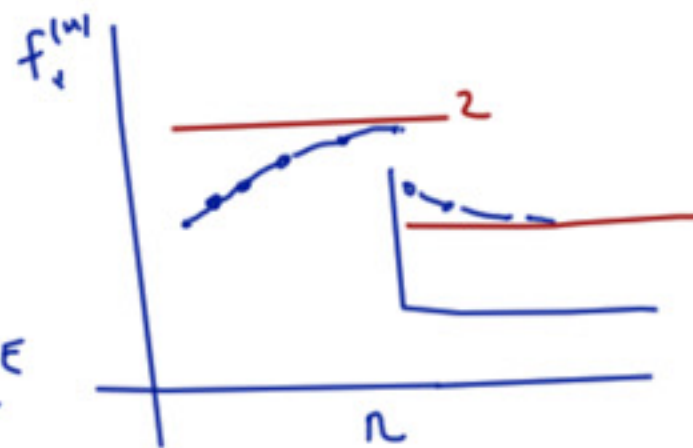


- MORE FACTORS?

SHORT RATE PROCESS

$$\begin{cases} z_t = z_{t+1} + \varepsilon_t \\ x_t = \phi x_{t+1} + \varepsilon_t \\ y_t^{(1)} = z_t + x_t \end{cases}$$

$$\begin{aligned} \rightarrow f_t^{(2)} &= E_t y_{t+1}^{(1)} = z_t + \phi x_t \rightarrow \text{LEVEL + SLOPE} \\ f_t^{(3)} &= E_t y_{t+2}^{(1)} = z_t + \phi^2 x_t \\ &\dots \end{aligned}$$



REAL MODEL? LIKE THIS BUT MODEL M_{t+1} , $P_t^{(n)} = E_t(M_{t+1} \dots M_{t+n})$

BONDS & DISCRETE TIME VASICEK

• RESULT

$$y_t^{(1)} \cdot \delta = \phi (y_{t-1}^{(1)} \cdot \delta) + \varepsilon_{t,1}$$

$$f_t^{(2)} = \delta + \phi (y_t^{(1)} \cdot \delta) - \left[\frac{1}{2} + \gamma \right] \sigma_\varepsilon^2$$

$$f_t^{(3)} = \delta + \phi^2 (y_t^{(1)} \cdot \delta) - \left[\frac{1}{2} (1+\phi)^2 + \gamma (1+\phi) \right] \sigma_\varepsilon^2$$

$$f_t^{(4)} = \delta + \phi^3 (y_t^{(1)} \cdot \delta) - \left[\frac{1}{2} (1+\phi^3)^2 + \gamma (1+\phi^3) \right] \sigma_\varepsilon^2$$

$$f_t^{(n)} = \delta + \phi^{n-1} (y_t^{(1)} \cdot \delta) - \left[\frac{1}{2} \left(\frac{1-\phi^{2n-1}}{1-\phi} \right)^2 + \gamma \left(\frac{1-\phi^{2n-1}}{1-\phi} \right) \right] \sigma_\varepsilon^2$$

EXPECTATIONS

RISK PREMIUM + $\frac{1}{2} \sigma^2$ TERMS

$$y_t^{(n)} = \frac{1}{n} \left[y_t^{(1)} + f_{t+1}^{(2)} + f_{t+2}^{(3)} + \dots + f_{t+n}^{(n)} \right]; P_t^{(n)} = -n y_t^{(n)} \dots$$

• DERIVATION

$$\begin{cases} x_{t+1} \cdot \delta = \phi (x_t \cdot \delta) + \varepsilon_{t+1} \\ \ln M_{t+1} = M_{t+1} = x_t - \frac{1}{2} \lambda^2 \sigma_\varepsilon^2 - \lambda \varepsilon_{t+1} \end{cases}$$

$$P_t^{(n)} = E_t (M_{t+1} M_{t+2} \dots M_{t+n})$$

• APPROACH 1: BRUTE FORCE

$$P_t^{(1)} = E_t M_{t+1} = E_t e^{M_{t+1}} = E_t e^{-x_t - \frac{1}{2} \lambda^2 \sigma_\varepsilon^2 - \lambda \varepsilon_{t+1}} = e^{-x_t}$$

$P_t^{(1)} = -x_t$ $y_t^{(1)} = x_t!$ "BOND PRICES REVEAL LATENT STATE VARIABLES"
CAN START WITH $y_t^{(1)} = x_t$, THEN CHECK MODEL PRODUCES $y_t^{(1)}$

$$\begin{aligned} P_t^{(2)} &= E_t (M_{t+1} M_{t+2}) = E_t (M_{t+1} P_{t+1}^{(1)}) = E_t e^{-x_t - \frac{1}{2} \lambda^2 \sigma_\varepsilon^2 - \lambda \varepsilon_{t+1}} \cdot e^{-x_{t+1}} \\ &= E_t e^{-x_t - \frac{1}{2} \lambda^2 \sigma_\varepsilon^2 - \lambda \varepsilon_{t+1} - \delta - \phi (x_t \cdot \delta) - \varepsilon_{t+1}} \\ &= E_t e^{-2\delta - (1+\phi)(x_t \cdot \delta) - \frac{1}{2} \lambda^2 \sigma_\varepsilon^2 - (1+\lambda) \varepsilon_{t+1}} \end{aligned}$$

$$P_t^{(2)} = -2\delta - (1+\phi)(x_t \cdot \delta) + \left(\frac{1}{2} + \gamma \right) \sigma_\varepsilon^2$$

$$f_t^{(2)} = P_t^{(1)} \cdot P_t^{(2)}$$

①

BONDS 9: OTHER APPROACHES

- "SOLVE" DISCOUNT FACTOR/INTEGRATE"

$$P_t^{(n)} = E_t(M_{t+n} M_{t+2}) = E_t \left[e^{-x_t - \frac{1}{2} x_t^2 \sigma_\varepsilon^2 - \lambda \varepsilon_{t+1} - \overbrace{[\delta + \phi(x_t - \delta) + \varepsilon_{t+1}] - \frac{1}{2} \lambda^2 \sigma_\varepsilon^2}_{\text{HAVE FUN?}} - \varepsilon_{t+2}} \right] \dots$$

- "DERIVE A DIFFERENCE/DIFFERENTIAL EQUATION FOR PRICE AS FUNCTION OF STATE VARIABLES"

• GUESS $P_t^{(n)} = A_n - B_n(x_t - \delta)$

• USE PRICING TO $P_t^{(n-1)} \rightarrow P_t^{(n)}$ $P_t^{(n)} = E_t(M_{t+n} P_{t+n}^{(n)})$

$$A_n - B_n(x_t - \delta) = \log E_t \exp \left[-x_t - \frac{1}{2} x_t^2 \sigma_\varepsilon^2 - \lambda \varepsilon_{t+1} + A_{n-1} - B_{n-1}(x_{t+1} - \delta) \right]$$

$$= \underbrace{-\delta + A_{n-1}}_{\text{red}} - \underbrace{(1 + B_{n-1}\phi)(x_t - \delta)}_{\text{green}} + \underbrace{(B_{n-1}\lambda + \frac{1}{2} B_{n-1}^2 \sigma_\varepsilon^2)}_{\text{red}}$$

$$\Rightarrow B_n = 1 + \phi B_{n-1} \quad ; \quad A_n = -\delta + A_{n-1} + (B_{n-1}\lambda + \frac{1}{2} B_{n-1}^2 \sigma_\varepsilon^2)$$

$$B_0 = 0 \quad B_1 = 1 \quad B_2 = 1 + \phi \quad A_0 = 0 \quad A_1 = -\delta \quad A_2 = -2\delta + (\lambda + \frac{1}{2}) \sigma_\varepsilon^2$$

$$B_3 = 1 + \phi + \phi^2 \dots$$

$$A_3 = -3\delta + [(1 + \phi)\lambda + \frac{1}{2}(1 + \phi)^2] \sigma_\varepsilon^2 \dots$$

$$B_n = \frac{1 - \phi^n}{1 - \phi}$$

RISK-NEUTRAL APPROACH

$$f_t^{(n)} = (1 - \phi^{n-1})i + \phi^{n-1} y_t^{(1)} - \left[\frac{1}{2} \left(\frac{1 - \phi^{n-1}}{1 - \phi} \right)^2 \lambda + \left(\frac{1 - \phi^{n-1}}{1 - \phi} \right) \right] \sigma_\varepsilon^2$$

$$f_t^{(n)} = (1 - \phi^{n-1}) \underbrace{\left[\delta - \frac{\lambda}{1 + \phi} \sigma_\varepsilon^2 \right]}_{\delta^*} + \phi^{n-1} y_t^{(1)} - \left[\frac{1}{2} \left(\frac{1 - \phi^{n-1}}{1 - \phi} \right)^2 \right] \sigma_\varepsilon^2$$

- THE CROSS-SECTIONAL MODEL IS THE SAME AS A MODEL WITH $\lambda = 0$ RISK NEUTRAL + DISTORTED MEAN $\delta^* = \delta - \frac{\sigma_\varepsilon^2 \lambda}{1 - \phi}$

$$x_{t+1} = \delta^* + \phi(x_t - \delta^*) + \varepsilon_{t+1}$$

$$M_{t+1} = -x_t$$


$$y_t^{(1)} = x_t$$

$$P_t^{(2)} = \log e^{-x_t - x_{t+1}} = \log e^{x_t \cdot \delta^* - \phi(x_t - \delta^*) - \varepsilon_{t+1}}$$

$$= -2\delta^* - (1 + \phi)(x_t - \delta^*) + \frac{1}{2} \sigma_\varepsilon^2$$

ALGEBRA IS EASIER!

10. COMMENTS ON SINGLE FACTOR VASICEK

LIKE EH: "SHORT RATE PROCESS" (FACTOR PROCESS) + "SINGLE FACTOR MODEL"
 ↓ (CROSS SECTION) 

↓ (HOW YIELDS MOVE OVERTIME FORECASTS)

RISK PREMIUM IN MODEL. SIGN OF λ : CAN GO EITHER WAY
 "EXPECTATIONS HYPOTHESIS + RISK NEUTRALITY" $\frac{1}{2}\sigma^2$. SMALL $\sigma \approx 0.01$

EXAMPLE GRAPH
 RISK-NEUTRAL: λ IS NOT NEEDED TO DESCRIBE CROSS SECTION.
 λ CONNECTS CROSS SECTION ("TOO HIGH" δ) WITH TIME SERIES FORECASTS.

→ "RISK NEUTRAL" USED FOR MANY PRICING CASES.
 → BE CAREFUL!! $\delta \neq \delta!$ [$\phi \neq \phi!$]

$M_{t+1} > 0 \rightarrow$ "ARBITRAGE FREE" MODEL PRICE $P = E(M_{t+1} X_{t+1}^c)$
 W/O ARBITRAGE (AT LEAST)
 BLACK-SCHOLES FOR BOND OPTIONS

3

TO DO
 MORE FACTORS: $X_t = \begin{bmatrix} X_t^1 \\ X_t^2 \\ \vdots \\ X_t^k \end{bmatrix}$ ← VECTOR

TIME-VARYING RISK PREMIUMS [FAMA-BLISS]

$$X_{t+n} - M = \Phi(X_t - M) + \Sigma_{t+n} \quad E(\Sigma_{t+n} \Sigma_{t+1}') = V$$

$$\log M_{t+n} = M_{t+n} = -\delta' X_{t+n} - \frac{1}{2} \lambda_t' V^{-1} \lambda_t - \lambda_t' \Sigma_{t+n}$$

$$\lambda_t = \lambda_0 + \Lambda_1 X_t$$

$$\Rightarrow M^* = M - V \lambda_0 \quad \Phi^* = \Phi - V \Lambda_1$$

$$f_t^{(n)} = \dots + \delta_1' \underbrace{\Phi^{*n-1}}_{\text{FIT SHAPE}} X_t$$

MORE... EASIER IN CONT. TIME.

BONDS II: CONTINUOUS TIME TERM STRUCTURE

- SAME IDEAS!
- SIMPLE CASE: VASICEK.

[$r = x$]
 $f(n, r) = \frac{\partial P(n, r)}{\partial n}$

$$\begin{cases} dr_t = \phi(F - r_t)dt + \sigma_r dz_r = \mu_r(r_t)dt + \sigma_r dz_r \\ \frac{dn_t}{n} = -r_t dr_t - \sigma_n dz_n \end{cases}$$

RESULT

$$f(n, r) \cdot f_+^{(r)} = \underbrace{\bar{r} + e^{-\phi n}(r - \bar{r})}_{\text{expectations}} - \underbrace{\left[\frac{1}{2} \frac{\sigma_r^2}{\phi^2} (1 - e^{-2\phi n})^2 + \sigma_n \frac{\sigma_r}{\phi} (1 - e^{-\phi n}) \right]}_{\text{variance, } 1/2 \sigma^2}$$

APPROACH - GUESS $P(n, r)$. FIND PDE FOR $P(n, r)$ FROM INSTANTANEOUS PRICING.

PRICING → DE

$$E_r(dR_t) - r_t dr_t = -E_r\left(\frac{dn_t}{n} \frac{dP}{P}\right) \quad (A)$$

Ito

$$dP(n, r) = \left[\frac{\partial P}{\partial r} \mu_r + \frac{1}{2} \frac{\partial^2 P}{\partial r^2} \sigma_r^2 \right] dt + \frac{\partial P}{\partial n} \sigma_n dz_n \quad (B)$$

DEFINITION OF $P(n, r)$
 ↑
 TIME TO GO
 NOT TIME OF MATURITY

(4)

(B) → (A)

$$\frac{\partial P}{\partial t} \mu_r + \frac{1}{2} \frac{\partial^2 P}{\partial r^2} \sigma_r^2 - \frac{\partial P}{\partial n} - rP = \frac{\partial P}{\partial r} \sigma_r \sigma_n$$

$$\frac{\partial P(n, r)}{\partial n} = \underbrace{(\mu_r - \sigma_r \sigma_n)}_{\mu} \frac{\partial P(n, r)}{\partial r} + \frac{1}{2} \frac{\partial^2 P(n, r)}{\partial r^2} \sigma_r^2 - rP(n, r)$$

- BOND PDE. LINE BLACK-SCHOLES PDE. $P(0, n) = 1$, SOLVE (NUMERICALLY) FOR n .
 $\rightarrow P(n=0, r)$

- GUESS $P(n, r) = e^{A(n) - B(n)r}$ FIND $A(n) | B(n)$

$$\frac{\partial P}{\partial r} = -B(n)P \quad ; \quad \frac{\partial^2 P}{\partial r^2} = B(n)^2 P \quad ; \quad \frac{\partial P}{\partial n} = [A'(n) - B'(n)r]P$$

$$\Rightarrow \underbrace{A'(n)} - \underbrace{B'(n)r} \cdot \cdot \left[\underbrace{\phi(r - r)}_{\phi} - \underbrace{\sigma_r \sigma_n} \right] \underbrace{B(n)} + \frac{1}{2} \underbrace{\sigma_r^2}_{\phi^2} \underbrace{B(n)^2} - r$$

$$B'(n) = 1 + \phi B(n) \quad ; \quad A'(n) = [\phi r - \sigma_r \sigma_n] B(n) + \frac{1}{2} \sigma_r^2 B(n)^2$$

* TURNED STOCHASTIC DIFFERENTIAL EQUATION + EXPECTATION $P(n, r) = E_r[\Lambda_{n,0} | \mathcal{F}_t]$ [HARD, BUT OK NUMERICALLY BY MONTE CARLO, SIMULATE r_t TURNED]
 INTO PARTIAL DIFFERENTIAL EQUATION [HARD, BUT OK NUMERICALLY]
 INTO ORDINARY DIFFERENTIAL EQUATION [REALLY FAST IF NUMERICAL; EASY HERE BY HAND]

$\frac{dB(n)}{dn} = 1 + \phi B(n) ; \int \frac{1}{1 + \phi B} dB = \int dn ; -\frac{1}{\phi} \log(1 + \phi B) = n ; B_n = \frac{1}{\phi} (1 - e^{-\phi n}) \leftrightarrow B(n) = \frac{1 - e^{-\phi n}}{1 - \phi} ; A(n)$ similarly.

12. CONTINUOUS TIME DISCUSSION

- 'PAIR PROCESS' (THE) + "FACTOR MODEL" (CROSS SECTION)
- EXPECTATIONS + RISK PREMIUM, $\frac{1}{2}\sigma^2$ TERMS

• RISK NEUTRAL TRICK $\bar{r} = \bar{r} - \frac{\sigma_n \sigma_r}{\phi}$

$$\Rightarrow f(t, r) = F' - e^{-\int_t^T (r - \bar{r})} - \frac{1}{2} \frac{\sigma_r^2}{\phi^2} (1 - e^{-\int_t^T})^2$$

PEEPER $\mu_r = \sigma_r \sigma_n$; $d\bar{r} = \sigma_r \sigma_n$ ONLY PLACE σ_n ENTERS

- SIMPLIFIES ALGEBRA, BUT DON'T FIT μ_r OR \bar{r} FROM TIME SERIES!
AP π

- MORE REALISTIC: SAME STRUCTURE, IDEAS.

- KEEP $r_t > 0$; SOME (RANDOM) VOLATILITY: CIR
 $dr_t = \phi(\bar{r} - r_t) + \sqrt{r_t} \sigma_r dz_t$

- MULTIPLE FACTORS: $r \rightarrow x$, A VECTOR
- "MULTIFACTOR AFFINE"

$$dx_t = \phi(x_t - \bar{x}_t) dt + \Sigma dz_t$$

$$r_t = \delta_0 - \delta_1' x_t$$

$$\frac{dn}{n} = -r_t dt - b_n' dz_t$$

$$dz_i = \sqrt{\alpha_i + \beta_i x_t} dz_{2t}$$

\rightarrow STILL ? $(n_t, x_t) = e^{-A(-1 - B_1 + 1)x_t}$
 $A(\cdot) B(\cdot)$ FAST DOE

- "UNSPANNED VARIABLES"

x_t , STOCHASTIC VAR. THAT IS NOT REPAIRED BY B AND P AND C.

- POINT OF ALL THIS

(WHICH MODEL - WHICH PURPOSE)

- CROSS SECTION - PRICE WEIRD BONDS, OPTIONS
- FIND PRICING ANOMALIES, "LIQUIDITY"

- FIND TRADING STRATEGIES

- FORECAST RATES

- INTERPRET CURVE \rightarrow EXPECTED REAL, π , PREMIUM?

