



SOCIETY OF ACTUARIES

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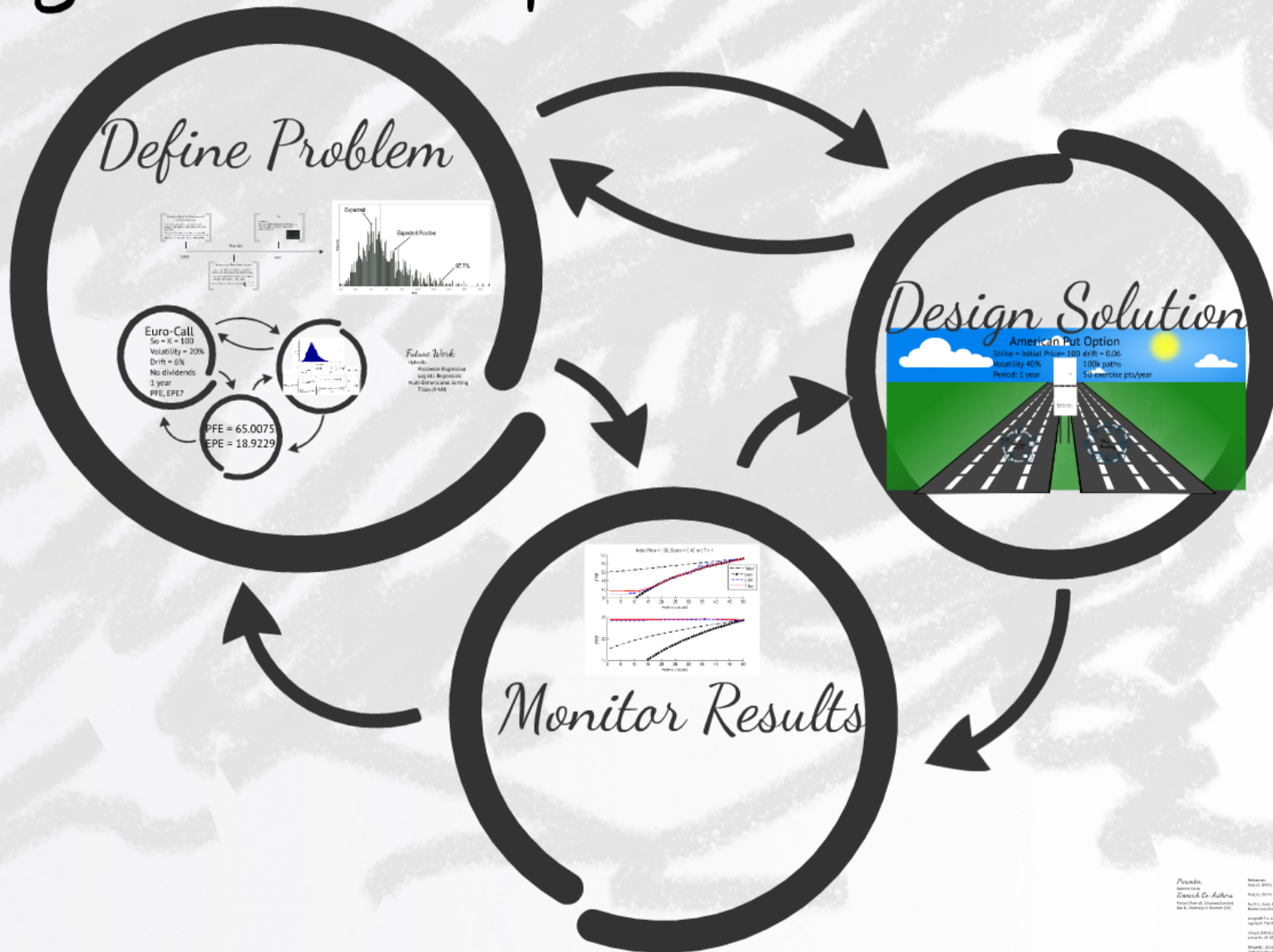
EXTERNAL FACTORS:

- Regulatory Environment
- Government and Judicial context
- Physical Environment
- Economic and Social Environment
- Industry and Business Environment

Shepherd (2008)



Pricing Risk through Simulation: Revisiting Tilley Bundling and Least Squares Monte Carlo Methods



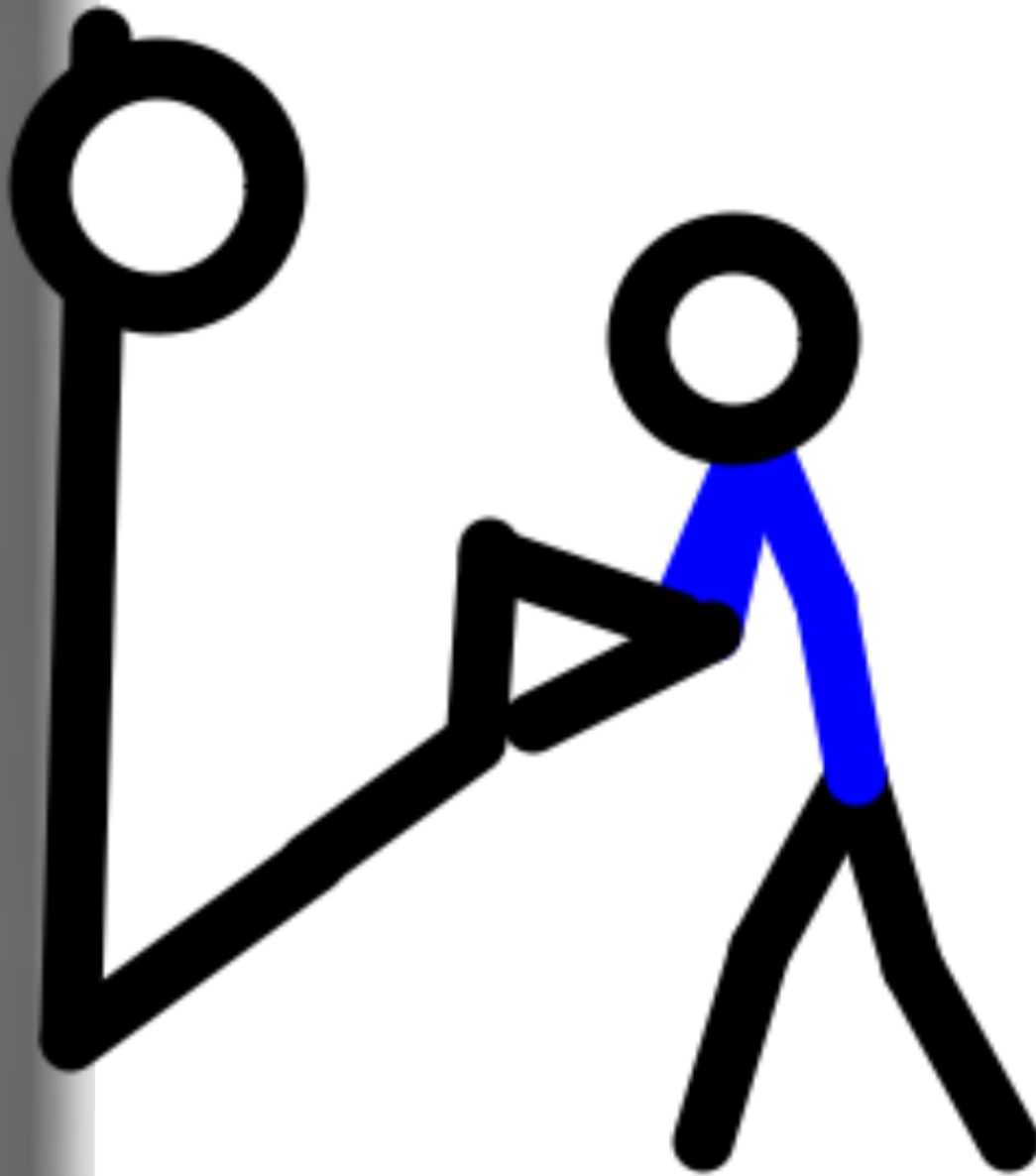
Presented at the 2015 International Conference on Financial Engineering and Applied Finance (ICFEAF) in London, UK. This work is based on the research of the author(s) and is not necessarily endorsed by the author(s). Copyright © 2015, All Rights Reserved. This work is licensed under a Creative Commons Attribution-NonCommercial-ShareAlike 4.0 International License. For more information, please visit: <http://creativecommons.org/licenses/by-nc-sa/4.0/>



EXTERNAL FACTORS:

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 - Government and judicial context
 - Physical Environment
 - Economic and Social Environment
 - Industry and Business Environment
- Sheperd (2006)

Regulatory Environment
Government and judicial context
Physical Environment
Economic and Social Environment
Industry and Business Environment
Sheperd (2006)



Professionalism

Simulation Models for Derivatives with early exercise features

"Monte Carlo simulation can only be used for European-style options." (Hull, 1993 as cited by Fu et al, 2001)

"The goal of this paper is to dispel the prevailing belief that American style options cannot be valued efficiently in a simulation model." (Tilley, 1993)

1993

90s-00s

Development of Monte Carlo techniques

Longstaff and Shwartz develop a simple least-squares approach. (Longstaff and Shwartz, 2001)

"Monte Carlo simulation is well suited to valuing path-dependent options ..." (Hull, 2012)

Fu et al (2001) consider 3 approaches

- semi backward induction algorithm
- perturbation corp. exercise time
- finding efficient, upper and lower bounds

Now

Risk Metrics
This presentation: Expected Positive Exposure,
Potential Future Exposure through Tilley
Bundling and LSM



now

Simulation Models for Derivatives with early exercise features

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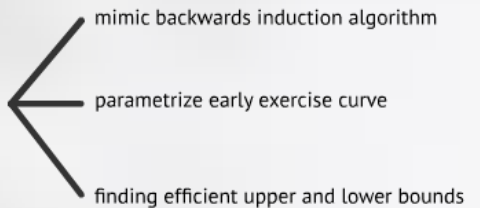
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Fu et al (2001) consider 3 approaches



- mimic backwards induction algorithm
- parametrize early exercise curve
- finding efficient upper and lower bounds

- mimic backwards induction algorithm
- parametrize early exercise curve
- finding efficient upper and lower bounds

Now

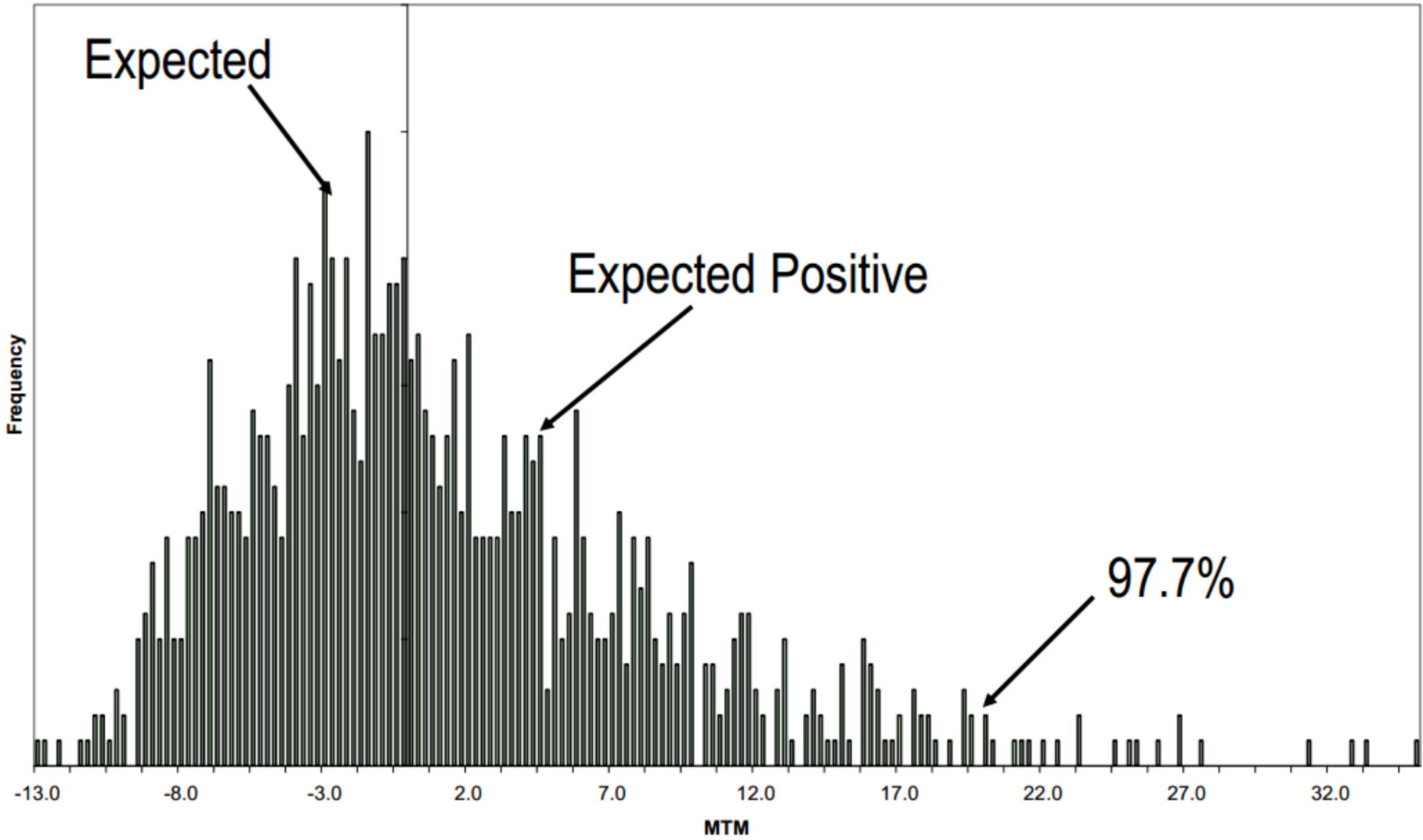
Risk Metrics

This presentation: Expected Positive Exposure,
Potential Future Exposure through Tilley
Bundling and LSM



- Basel II/III - Defines EPE, PFE
- Technological progress
- Convergence of financial institutions:
risk still there when in the money

JET 390124 4" HD



Euro-Call

$$S_0 = K = 100$$

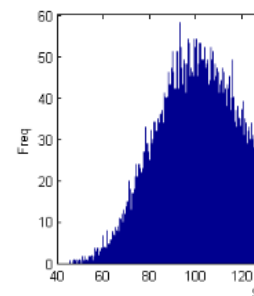
$$\text{Volatility} = 20\%$$

$$\text{Drift} = 6\%$$

No dividends

1 year

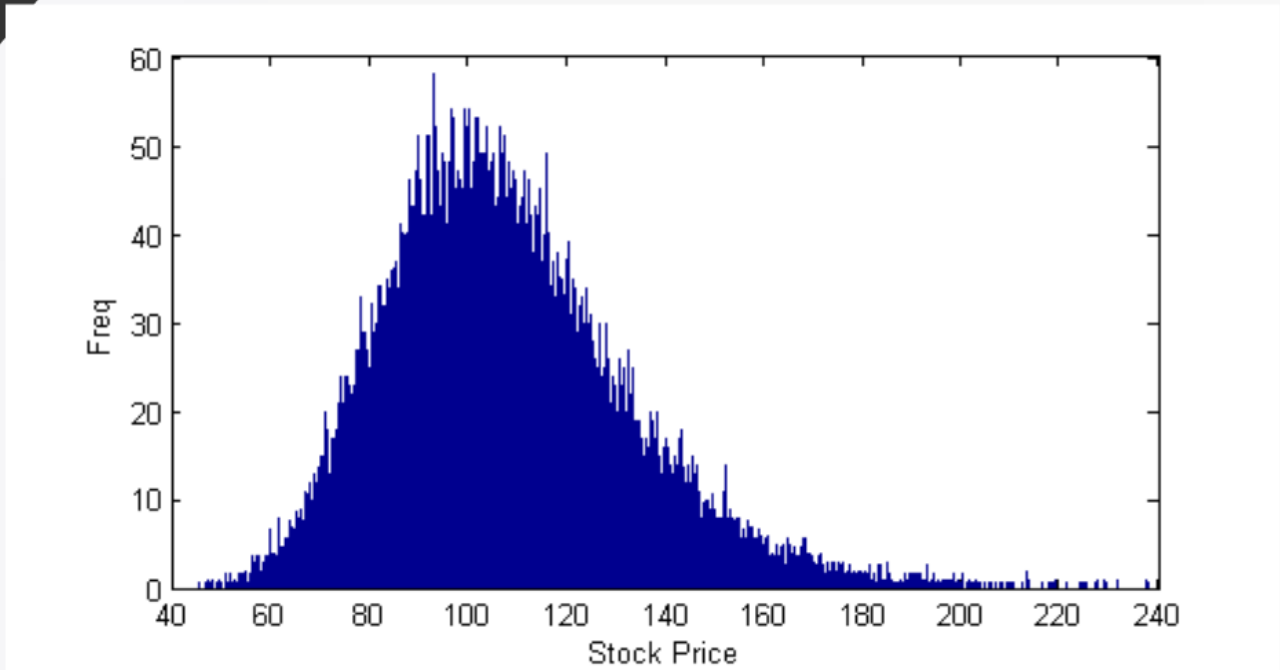
PFE, EPE?



$$EPE = e^{-0.06} \frac{\int_{100}^{\infty} \frac{1}{x\sigma\sqrt{2\pi}} \exp\left[-\frac{(\ln x - 0.06 - 0.2^2/2)^2}{2(1 - 0.2^2)}\right] dx}{\int_{100}^{\infty} \frac{1}{x\sigma\sqrt{2\pi}} \exp\left[-\frac{(\ln x - 0.06 - 0.2^2/2)^2}{2(1 - 0.2^2)}\right] dx}$$

$$\text{PFE} = 65.0075$$

$$\text{EPE} = 18.9229$$



$$\int_0^{e^{0.06}PFE_{0.99}+100} \frac{1}{x\sigma\sqrt{2\pi}} \exp \left[-\frac{\left(\log(x) - \log(100) - \left(\mu - \frac{\sigma^2}{2}\right)\right)^2}{2\sigma^2} \right] dx = 0.99$$

$$EPE = e^{-0.06} \left(\frac{\int_{100}^{\infty} \frac{1}{\sigma\sqrt{2\pi}} \exp \left[-\frac{\left(\log(x) - \log(100) - \left(\mu - \frac{\sigma^2}{2}\right)\right)^2}{2\sigma^2} \right] dx}{\int_{100}^{\infty} \frac{1}{x\sigma\sqrt{2\pi}} \exp \left[-\frac{\left(\log(x) - \log(100) - \left(\mu - \frac{\sigma^2}{2}\right)\right)^2}{2\sigma^2} \right] dx} - 100 \right)$$

PFE = 65.0075

EPE = 18.9229

Signs on the Road

American Put Option

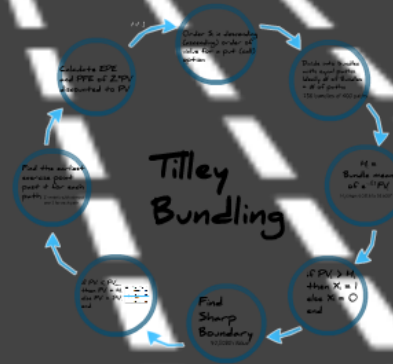
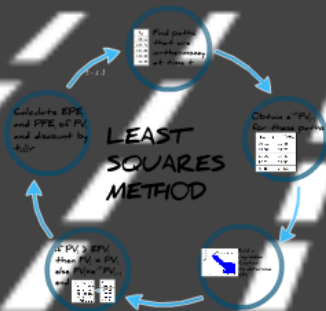
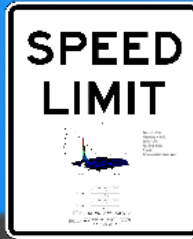
Strike = Initial Price = 100 drift = 0.06

Volatility 40%

Period: 1 year

100k paths

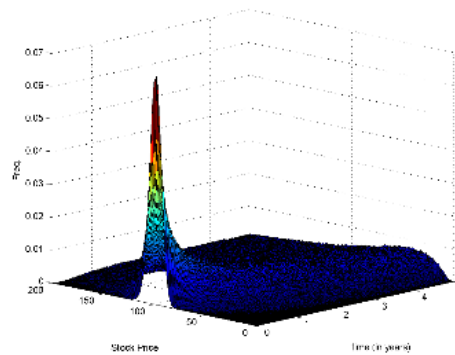
50 exercise pts/year



MINIMUM SPEED

Metrics for a European
Option ending at time t

SPEED LIMIT



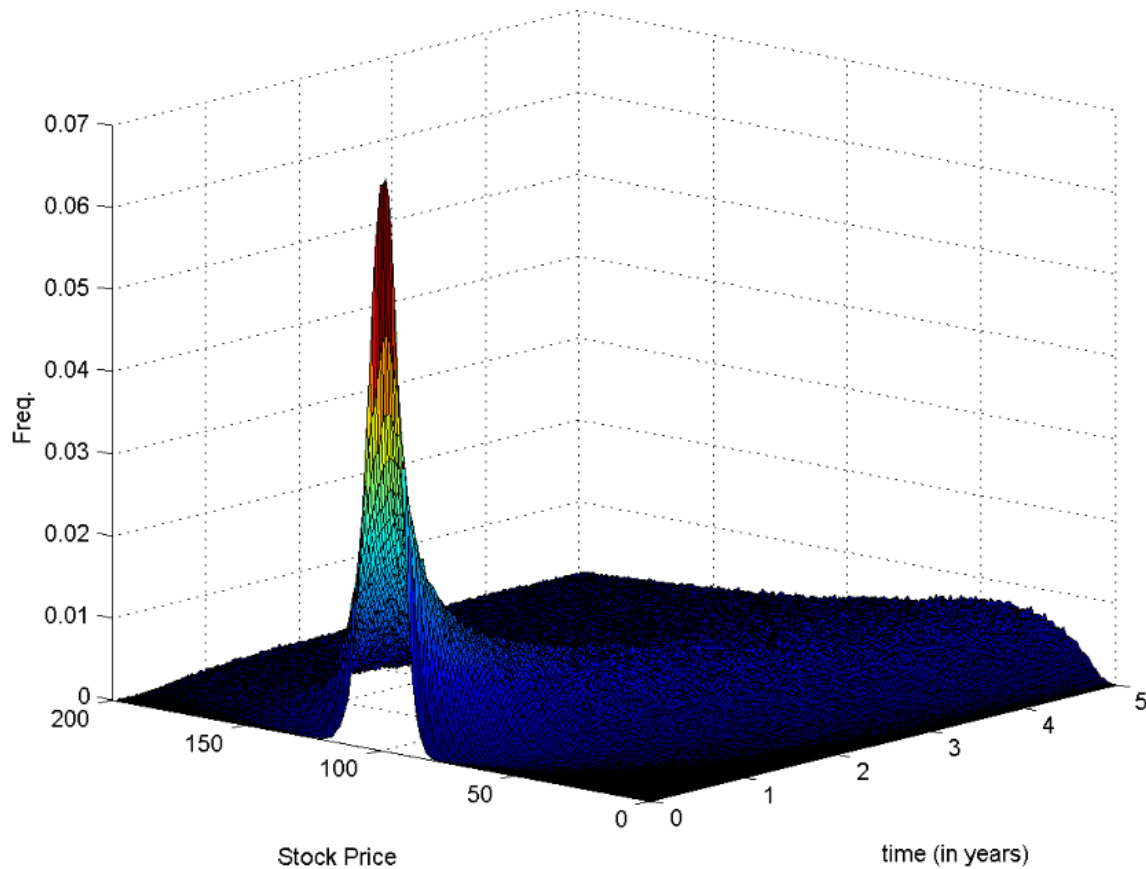
$S_0 = K = 100$
 Volatility = 40%
 Drift = 6%
 No dividends
 5 year
 50 exercise times/ year

$$F_2(x, t, a) = \int_0^a \frac{\exp \left[-\frac{(\log(x) - \log(S_0) - (\mu - \frac{\sigma^2}{2})t)^2}{\sigma^2 t} \right]}{\sigma \sqrt{\pi t}} dx$$

$$F_3(x, t, a) = \int_0^a \frac{\exp \left[-\frac{(\log(x) - \log(S_0) - (\mu - \frac{\sigma^2}{2})t)^2}{\sigma^2 t} \right]}{\sigma \sqrt{\pi t}} dx$$

$$\sum_{t=k}^{50} F_2(x, t, 100 - PFE_{0.99} e^{0.0012t}) = 0.01(51 - k)$$

$$EPE_{k^{th} \text{ time}} = \frac{\sum_{t=k}^{50} e^{-0.0012t} F_2(x, t, 100) \left(100 - \frac{F_3(x, t, 100)}{F_2(x, t, 100)} \right)}{(51 - k) \sum_{t=k}^{50} F_2(x, t, 100)}$$



$S_0 = K = 100$
 Volatility = 40%
 Drift = 6%
 No dividends
 5 year
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$$F_0(x, t, \sigma) = \int_a^x \frac{\exp \left[-\frac{\left(\log(x) - \log(S_0) - \left(\mu - \frac{\sigma^2}{2} \right) \frac{t}{50} \right)^2}{\sigma^2 \frac{t}{25}} \right]}{\sigma \sqrt{\frac{t}{25}}} dv$$

$$F_2(x, t, a) = \int_0^a \frac{\exp \left[-\frac{(\log(x) - \log(S_0) - (\mu - \frac{\sigma^2}{2}) \frac{t}{50})^2}{\sigma^2 \frac{t}{25}} \right]}{x \sigma \sqrt{\pi \frac{t}{25}}} dx$$

$$F_3(x, t, a) = \int_0^a \frac{\exp \left[-\frac{(\log(x) - \log(S_0) - (\mu - \frac{\sigma^2}{2}) \frac{t}{50})^2}{\sigma^2 \frac{t}{25}} \right]}{\sigma \sqrt{\pi \frac{t}{25}}} dx$$

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S_{25}
98.12
144.72
131.41
113.38
54.38

Find paths that are in-the-money at time t

$t = t-1$

Calculate EPE_t and PFE_t of FV_t and discount by $+dtr$

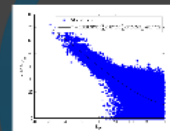
LEAST SQUARES METHOD

Obtain $e^{-dtr} FV_{t+1}$ for these paths

S_{25}	$e^{-0.0012} FV_{26}$
98.12	17.45
54.38	31.21
69.27	33.47
80.24	22.91
83.09	28.66

if $PV_t > EFV_t$
 then $FV_t = PV_t$
 else $FV_t = e^{-dtr} FV_{t+1}$
 end

S_{25}	PV_{25}	EFV_{25}	PV_{26}	FV_{26}
98.12	1.88	11.10	1.88	17.45
144.72	0.00	31.39	0.00	31.21
131.41	0.00	16.51	0.00	33.47
113.38	0.00	8.72	0.00	22.91
54.38	45.62	45.49	45.62	45.62



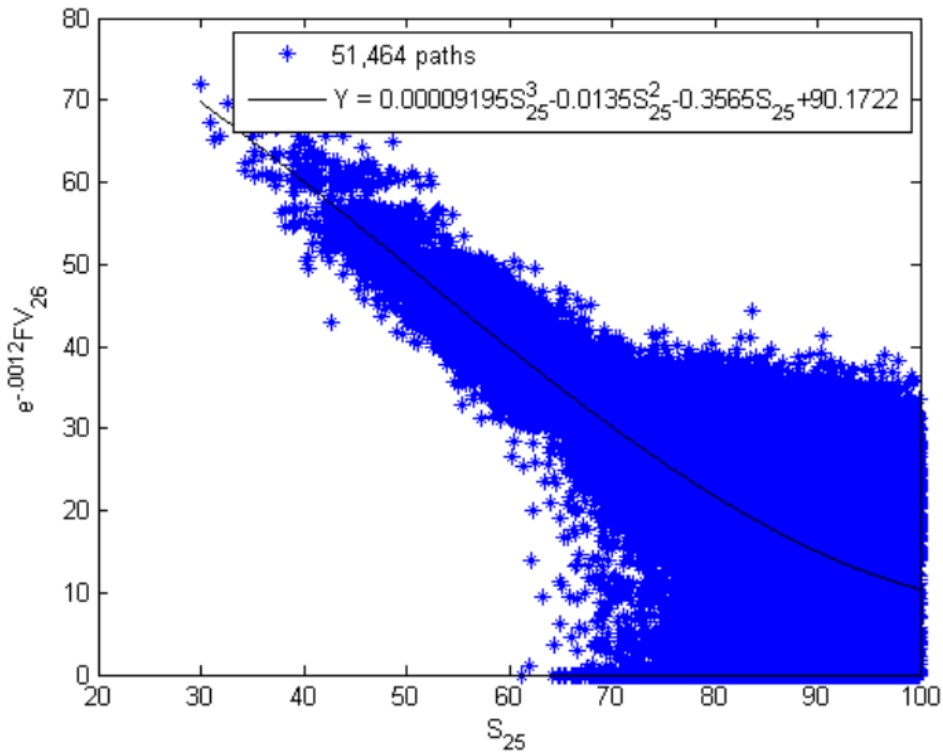
Build a regression function to determine EFV_t

S_{25}
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Find paths
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Obtain $e^{-dt}r FV_{t+1}$
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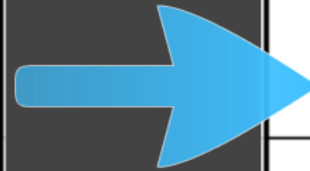
end

S_{25}	PV_{25}	EFV_{25}
98.12	1.88	11.10
144.72	0.00	31.39
131.41	0.00	16.51
113.38	0.00	8.72
54.38	45.62	45.49



PV_{25}	FV_{25}
1.88	17.45
0.00	31.21
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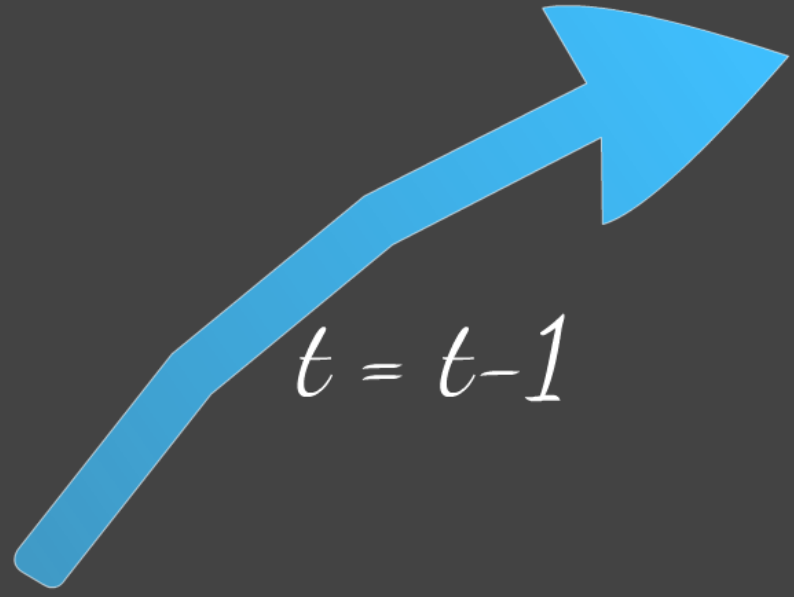
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PV_{25}	FV_{25}
1.88	17.45
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Calculate EPE_t
and PFE_t of FV_t
and discount by
 $+dtr$

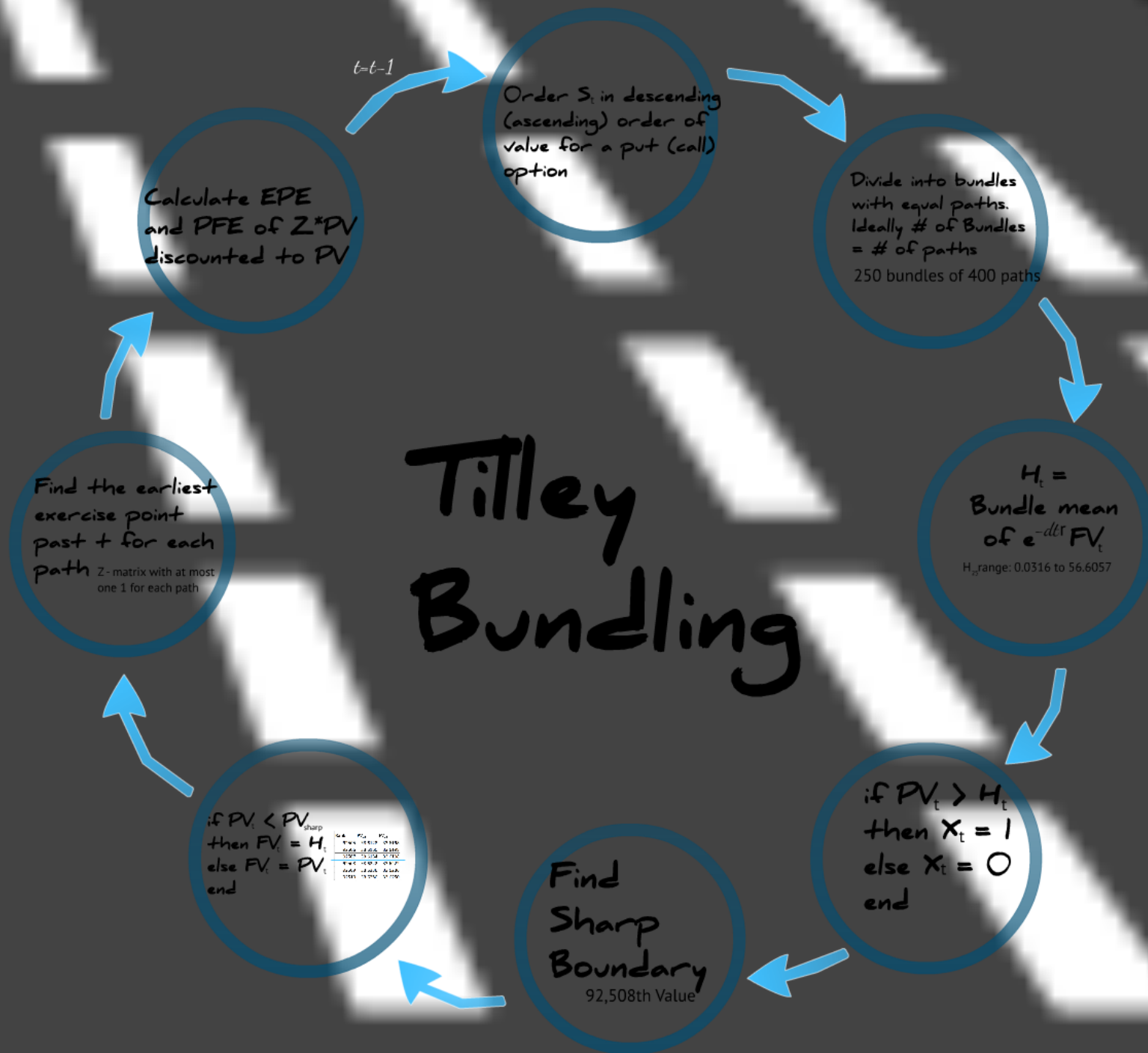
S_{25}
98.12
144.72
131.41
113.38
54.38



Calculate EPE_t
and PFE_t of FV_t
and discount by

LEA

Tilley Bundling



Calculate EPE and PFE of $Z \cdot PV$ discounted to PV

$t=t-1$

Order S_i in descending (ascending) order of value for a put (call) option

Divide into bundles with equal paths. Ideally # of Bundles = # of paths
250 bundles of 400 paths

$H_t =$
Bundle mean of $e^{-dt \cdot r} FV_t$
 H_t range: 0.0316 to 56.6057

if $PV_t > H_t$
then $X_t = 1$
else $X_t = 0$
end

Find Sharp Boundary
92,508th Value

if $PV_t < PV_{sharp}$
then $FV_t = H_t$
else $FV_t = PV_t$
end

$t=4$	PV_t	PV_{sharp}
0.00	0.00	0.00
0.01	0.01	0.01
0.02	0.02	0.02
0.03	0.03	0.03
0.04	0.04	0.04
0.05	0.05	0.05
0.06	0.06	0.06
0.07	0.07	0.07
0.08	0.08	0.08
0.09	0.09	0.09
0.10	0.10	0.10

Find the earliest exercise point past t for each path
Z-matrix with at most one 1 for each path

Order S_t in descending
(ascending) order of
value for a put (call)
option

Divide into bundles
with equal paths.
Ideally # of Bundles
= # of paths

250 bundles of 400 paths

$H_t =$
Bundle mean
of $e^{-dt} r FV_t$

H_{25} range: 0.0316 to 56.6057

if $PV_t > H_t$

then $X_t = 1$

else $X_t = 0$

end

Find
Sharp
Boundary

92,508th Value

if $PV_t < PV_{\text{sharp}}$
then $FV_t = H_t$
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end

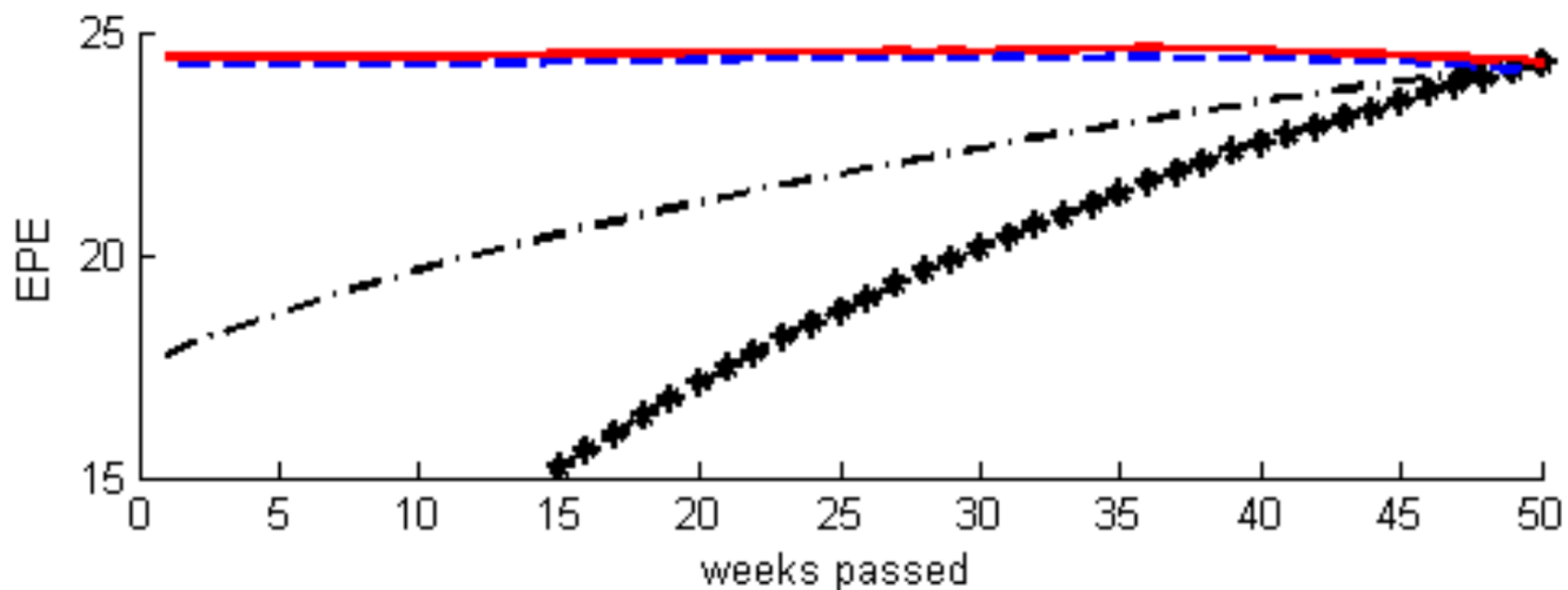
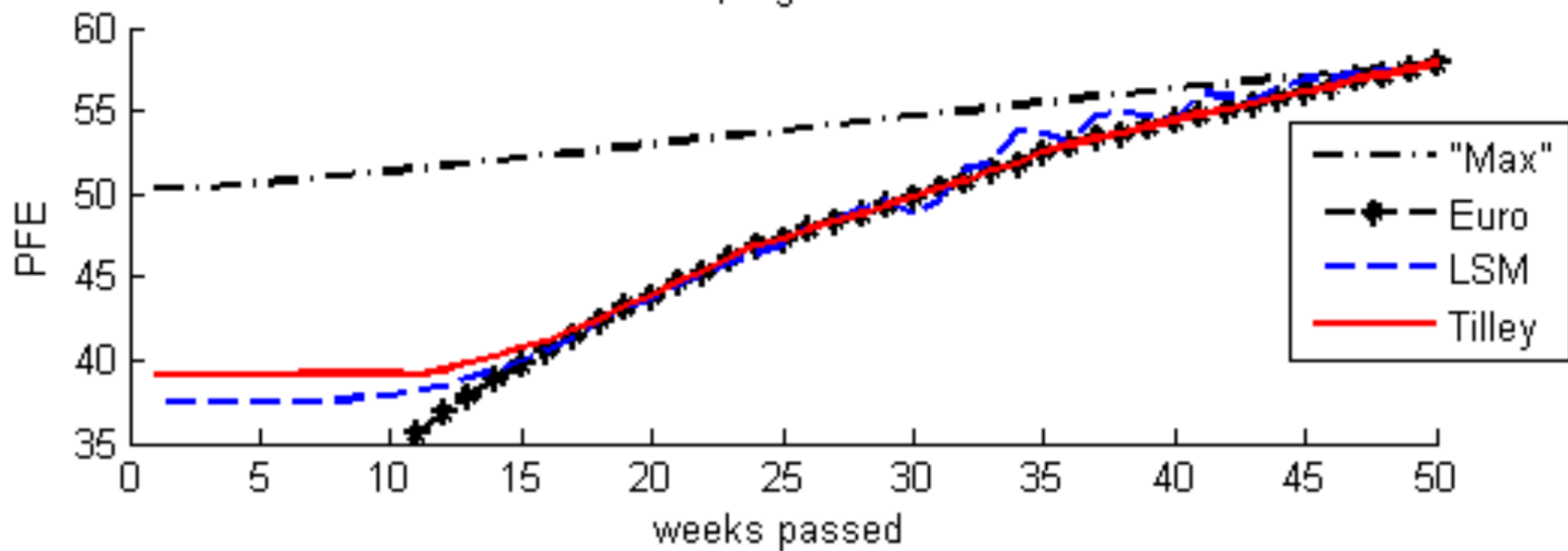
Rank	PV_{25}	FV_{25}
92505	33.6128	33.6198
92506	33.6160	33.6198
92507	33.6184	33.6198
92508	33.6222	33.6222
92509	33.6236	33.6236
92510	33.6250	33.6250

Find the earliest
exercise point
past t for each
path

Z - matrix with at most
one 1 for each path

Calculate EPE
and PFE of Z^*PV
discounted to PV

Initial Price = 100, Sigma = 0.40 and T = 1



Future Work:

Hybrids:

Piecewise Regression

Logistic Regression

Multi-Dimensional Sorting

Tilley (K-NN)

Presenter:

Dominic Cortis

Research Co-Authors:

Rickard Branvall, Citigroup (London)

Juxi Li, University of Leicester (UK)

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