A new financial calculus Introduction or ⊞ Financial the problem of computing hedging errors in models incomplete markets  $\boxplus$  Implementation 大 Stefan Dirnstorfer 2004-Feb-26

Introduction ⊞ Financial models ⊞ Implementation	Why a new financial notation The new operator notation has three advantages over prevailing financial calculus and the impact of the impact of the impact of the calculus straight forwardly induces the numerical procedure. Details of the portfolio and the intended hedging strategy can be expressed mathematically, including embedded options, even triggers and non-deta hedges. Business activities are integrated smoothly. The calculus separates process, portfolio and measure, to be defined by investment and sales, researchers and management.	Repetition of operator calculus Operator domain $O: (S_1 \rightarrow S_2) - (S_1 \rightarrow S_2) \qquad (1)$ Equivalence $O_1 \equiv O_2 \iff \forall V \in (S_1 \rightarrow S_2), : O_1 V = O_2 V \qquad (2)$ Associativity $(O_1 (O_2) V(s) = (O_1 (O_2 V))(s) \qquad (3)$ Addition $(O_1 + O_2) V = O_1 V + O_2 V \qquad (4)$	<b>Operator power</b> An operator power is a simplified way of writing multiple applications of one operator. For rational, functional and vector powers, the definition must fulfill the following properties: $\begin{pmatrix} O^{\mu} &= M & (b, c) \\ O^{\mu} &= O \\ O^{\mu} O^{\mu} &= O^{\mu+b} & a \in \mathbb{R} \lor b \in \mathbb{R} \ (b) \\ (O^{\mu})^{b} &= O^{\mu+b} \\ \end{pmatrix}$
	Why	Operators1	Operators2

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#### Why a new financial notation

The new operator notation has three advantages over prevailing financial calculus

- A numerical evaluation rule is implied. The operator calculus straight forwardly induces the numerical procedure.
- Details of the portfolio and the intended hedging strategy can be expressed mathematically, including embedded options, event triggers and non-delta hedges.
- Business activities are integrated smoothly. The calculus separates process, portfolio and measure, to be defined by investment and sales, researchers and management.

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#### Repetition of operator calculus

**Operator domain** 

$$O: (S_1 \to S_2) \to (S_1 \to S_2) \tag{1}$$

Equivalence

$$O_1 \equiv O_2 \iff \forall V \in (S_1 \to S_2), : O_1 V = O_2 V \qquad (2)$$

Associativity

$$(O_1 O_2)V(s) = (O_1(O_2 V))(s)$$
 (3)

Addition

$$(O_1 + O_2)V = O_1V + O_2V (4)$$

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**O**perator power

An operator power is a simplified way of writing multiple applications of one operator. For rational, functional and vector powers, the definition must fullfill the following properties:

$$O^{0} = Id$$

$$O^{1} = O$$

$$O^{a}O^{b} = O^{a+b} \qquad a \in \mathbb{R} \lor b \in \mathbb{R}$$

$$(6)$$

$$(O^{a})^{b} = O^{ab}$$

Where Id is the identity operator IdV = V.

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	Why	Operators1	Operators2



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#### **Financial Model**

A financial model is largely set up by three components: the process, the portfolio and a measure.

- The time **process** is a stochastic description of the path an observable variable can take.
- A <u>portfolio</u> describes a set of investments or the room for further trading activity.
- A <u>measure</u> is a computable property of the model. (E.g. expectations, distributions and parameters)

such that

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Suppose process and portfolio are each represented by an operator named  $\Theta$  and A. The measure is a function V,

Notation

$$(\Theta A)^M V(s) \tag{7}$$

becomes a unique and easy to evaluate answer to the question: What is the measure V of portfolio A under process  $\Theta$  over a horizon of length M, given a current economic state s.





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#### The time process

Let  $s_t$  be the time dependent vector of economic parameters and V a function.

$$\Theta : (\mathbb{R}^n \to \mathbb{R}^m) \to (\mathbb{R}^n \to \mathbb{R}^m)$$

$$\Theta V(s_t) = \mathbb{E}(V(s_{t+1}))$$
(8)

Applying  $\Theta$  to a measure V returns the expected future value of that measure.

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Time steps

Subdivisions of  $\Theta$  can be defined accordingly. This talk will use the terms  $\sqrt{\Theta}$  or  $\Theta^{\Delta t}$  to denote arbitrary time steps.

$$\Theta^{\Delta t} V(s_t) = \mathbb{E}(V(s_{t+\Delta t}))$$

**Note:** The discrete time notation is convenient, since it does not exclude, while not being limited to, processes which simply are not infinitely dividable, e.g. discrete state processes generated by matrices of transition probabilities or free form probability distributions.

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**Operator projection** 

The operator projection allows a partial definition of  $\Theta$ , i.e. its impact on only a subset of state variables. This is useful whenever the full information on the operator is either unknown or irrelevant.

$$\Theta|_{s}V(r_{t}, s_{t}) := \mathbb{E}(V(r_{t}, s_{t+1}))$$
(9)



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#### The shift operator

The operator  $T_x$  transforms a measure V by a deterministic growth in variable x.

**Definition** The operator definition declares the T operator by the transport equation with velocity  $\nu$  over a unit time interval.

$$\frac{dT_x^{\nu\epsilon}}{d\epsilon}V = \nu \frac{d}{dx}V \tag{10}$$

The variable  $\epsilon$  plays the role of the time variable in standard PDE theory.

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#### Example: Interest rate account

The following example models a cash account c that receives an interest rate dependent coupon  $cr_t$  in each period.

$$T_t V(t) = V(t+1)$$

$$T_c^{cr} V(c) = V(e^r c)$$
(11)

The shift  $T_c^{cr_t}$  pays the time dependend rate  $r_t$  and  $T_t$  increases a time variable t in each period.

$$(T_c^{cr_t} T_t)^M V(c) = (T_c^{cr_t} T_t)^{M-1} V(e^{r_t} c)$$
(12)  
=  $(T_c^{cr_t} T_t)^{M-2} V(e^{r_t} e^{r_{t+1}} c)$   
 $\cdots = V \left( c \prod_{p=0}^{M-1} e^{r_{t+p}} \right)$ 



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#### The blur operator

As soon as uncertainty enters the model we need a diffusion operator. The blur operator B solves the heat equation that has Brownian motion as source of randomness.

**Definition** The operator B solves the heat equation with diffusion coefficient  $\sigma^2/2$ . The size of  $\sigma$  determines the amount of uncertainty and was named **volatility** in financial terminology.

$$\frac{dB_x^{\sigma^2 \epsilon} V}{d\epsilon} = \frac{\sigma^2}{2} \frac{d^2}{dx^2} V \tag{15}$$



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#### Black & Scholes

The Black-Scholes model has Brownian motion as driving process, a european option in the portfolio and measures the fair premium for the option. Using the shift and blur operator, the time process  $\Theta_{BS}$  of this famous model can be rewritten.

$$\Theta_{BS} = e^{-r} T^{rS} B^{(\sigma S)^2} \tag{17}$$

The process is built from three **individual operations**: discounting  $e^{-r}$ , stock price drift  $T^{rS}$  and stock price diffusion  $B^{(\sigma S)^2}$ .

**Note:** All operators commute for constant  $\sigma$  and r

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#### Heath-Jarrow-Morton

The HJM model propagates a forward rate curve f(M) + r, whereas r is the rate change (initially 0).

$$\Theta_{HJM} = B_r^{\sigma^2} T_r^{\sigma\sigma^*} T_t \tag{18}$$

with

$$\sigma^* = \int_{t+\frac{1}{2}}^M \sigma dt' \tag{19}$$

whereas  $\sigma$  may depend on r, t and M.

**Note 1:** The integration starts with  $t + \frac{1}{2}$  instead of t due to the discrete time step.

Note 2: r is a dimension and not a variable. Non parallel curve shifts are possible:  $T_r^M V(1+r) = V(1+M+r)$ 

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#### Events

If two scenarios  $O_1$  and  $O_2$  occur with probabilities p and 1-p, their values can be combined linearly in the time process, since the expectation value  $\mathbb{E}$ , and therefore  $\Theta$ , is linear

$$\bigvee_{l \to p}^{p} O_{1} := p O_{1} + (1-p)O_{2}$$
(20)



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#### Correlation

The correlated diffusion can either be defined by a new PDE operator or on a transformed domain:

$$\Theta_{corr} = \underbrace{T_x^{-y} B_y^{\rho} T_x^{y}}_{1\text{-correlated}} B_x^{1-\rho} B_y^{1-\rho}$$
(21)

Correlations: 0, 0.5 and 0.8



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#### Convolution

Using the T operator we can formulate the convolution with a probability distribution  $\Phi$ :

$$V * \Phi'(x) = \int_0^1 T_x^{-\Phi^{-1}(u)} V du$$
 (22)

which is equivalent to the standard form:

$$\int_{0}^{1} T_{x}^{-\Phi^{-1}(u)} V(x) du = \int_{0}^{1} V(x - \Phi^{-1}(u)) du$$
$$= \int_{-\infty}^{\infty} V(x - u) \Phi'(u) du$$









#### Activity operator

The definition of an activity operator A is similar the process operator  $\Theta$ , with the only difference, that A has no expansion in time.

$$A : (\mathbb{R}^n \to \mathbb{R}^m) \to (\mathbb{R}^n \to \mathbb{R}^m)$$
(24)  
$$AV(s_{t^-}) = \mathbb{E}(V(s_{t^+}))$$

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# Time line

Consider a portfolio with three actions taking place at different points in time.



The operators for this activity are ordered strictly chronologically.

$$A_1 \Theta^{\Delta t} A_2 \Theta^{\Delta t} A_3 \tag{25}$$



Transfer	
A transfer of goods, assets or cash is performed by the $T$ operator.	т
$T_c^n$ (2	26)
Whereas $n$ is the number of units transferred into the account $c$ .	
1 ) V	Transfer A transfer of goods, assets or cash is performed by the $T$ perator. $T_c^n$ (2) Whereas $n$ is the number of units transfered into the account $c$ .





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#### Options

An option is defined by the alternatives among which can be selected and by the entity that does select. The option values are denoted as  $V_1$  and  $V_2$ . The deciding entity is chracterized by her utility operator U.



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#### Time continuous option

An American option is an option with continuous execution right. The option requires a time process that can be infinitely subdivided and an option that is executable after each infinitely small time step.

Assume a utility operator U and a selectable option X. The process is defined by a limit of finite timesteps.

$$\lim_{\Delta t \to 0} \left( \Theta^{\Delta t} \xrightarrow{\mathcal{U} \times \mathcal{T}^{\mathcal{U}}}_{\mathcal{T} \times \mathcal{T}} X \right)^{\frac{M}{\Delta t}}$$
(30)



(31)

(32)

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#### Example

Consider a portfolio manager who can buy an arbitrary amount of stocks at price S. The money is extracted from account c, while the stock is added to the portfolio h.

The optimal investment is:

$$(T_c^{-S} T_h)^\star \tag{33}$$

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#### Delta Hedge

Assume two state variables c for the amount of cash in our account and h, the number of stock in our depot.

In order to hedge a trading activity A, consider a counter investment  $A_H$  that buys  $-\Delta$  stocks.

$$A_H V = \left(T_c^{-S} T_h\right)^{-\Delta} V \tag{34}$$

Whereas  $\Delta$  is the sensitivity of measure V to stock price changes

$$\Delta = \frac{d}{dS}V\tag{35}$$

Now the following time continuous strategy is risk free:

$$\lim_{\Delta t \to 0} \left( A_H \Theta^{\Delta t} \right)^{\frac{M}{\Delta t}} A \tag{36}$$

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#### Best incomplete hedge

Assume that a hedging transaction  $A_H$  produces costs k relative to the traded volume  $|\Delta|$ .

$$A_H V = T_{\Delta}^{\star} T_c^{-k|\Delta|} \left(T_c^{-S} T_h\right)^{\Delta} V \tag{37}$$

With the optimization decision  $T^*$ , the heding activity  $A_H$  optimizes the portfolio with respect to utility U.







amount of cash, assets or goods that might be otherwise obtained under the economic state.

$$N : \mathbb{R}^n \to \mathbb{R}^+ \tag{39}$$

![](_page_44_Figure_1.jpeg)

#### Common Numeraires

**Currency numeraires** Currency or cash numeraires can be converted linearly into each other.

$$\boldsymbol{\in}(s) = c_{\boldsymbol{\ast}}(s)\boldsymbol{\boldsymbol{\$}}(s) = c_{\boldsymbol{\imath}}(s)\boldsymbol{\boldsymbol{\imath}}(s) \tag{40}$$

**Discounted numeraires** Discounted or time dependent numeraires include the value decay.

$$N = \in e^{\int_0^t f(u)du} \quad \text{or} \quad \in e^{\sum_{u=0}^{t-1} f(u)}$$
(41)

**Utility numeraires** Subjective utility may be any transformation of an existing numeraire.

$$\tilde{N} = u(N) \tag{42}$$

![](_page_45_Figure_1.jpeg)

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#### Sensitivity

Let the resulting information from a financial model be:

$$(\Theta A)^M V(s) \tag{43}$$

The sensitivity to can be computed for any influencing variable by differentiation.

$$\frac{d}{dp} \left(\Theta_p A_p\right)^M V_p(s_p) \tag{44}$$

#### The Greeks Introduction Common characterisations of a portfolio: Financial models $\blacksquare$ Process $\Delta = \frac{d}{dS} (\Theta A)^M N(s)$ $\square$ Portfolio Measure Ħ $\Gamma = \frac{d^2}{dS^2} (\Theta A)^M N(s)$ Sensitivity Distribution (Vega) $v = \frac{d}{d\sigma} (\Theta A)^M N(s)$ Stop time Parameter $-\theta = \frac{d}{d\tau} \Theta^{\tau} (\Theta A)^{M} N(s)$ $\rho = \frac{d}{dr} (\Theta A)^{M} N(s)$ fitting Implementation

(45)

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#### Example

The process is Heath-Jarrow-Morton  $\Theta_{HJM}$  with

$$\sigma = \frac{1\%}{10\% + M^2} \sqrt{f(M)}$$
(46)  
$$f(M) = 5\% + 0.5\% M + r$$

The portfolio is a bond combined with a short bermudean option:

$$PV = \left(\Theta T_c^{7\epsilon}\right)^2 \begin{pmatrix} T_c^{105\epsilon} V & T_c^{105\epsilon} V \\ T_c^{105\epsilon} V & \Theta T_c^{7\epsilon} \end{pmatrix}^8 T_c^{100\epsilon} V \quad (47)$$

As measure, we want to know the discounted cash on our account:

$$\boldsymbol{\in} = \exp\left(-\sum_{u=0}^{t-1} f(u)\right) \tag{48}$$

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#### Present value

Since the option is short (worst scenario is chosen) our portfolio value is striktly below all optional bonds.

Bond value:

 $B = (\Theta T^{7\epsilon})^{10} T^{100\epsilon}$ 

First execution value:  $E = (\Theta T^{7\epsilon})^2 T^{105\epsilon}$ 

![](_page_49_Figure_7.jpeg)

#### Distribution

The probability distribution is a measure V, created from a numeraire N and a variable x.

$$V = \binom{N}{1_{N>x}} \tag{49}$$

The utility operator U has to be adjusted to the new measure:

$$\tilde{U}V = UV_1 \tag{50}$$

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![](_page_51_Figure_1.jpeg)

#### Stop times

For stop time measure we use a stop condition C.

$$\tilde{V} = \begin{pmatrix} V \\ 1_{C(V)} \end{pmatrix} \tag{51}$$

The time process has to be adapted to test for the condition to be satisfied after each step.

$$\tilde{\Theta}\begin{pmatrix}V\\x\end{pmatrix} = \begin{pmatrix}\Theta V\\ O \\ O \\ O \\ O \\ X \end{pmatrix}$$
(52)

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![](_page_53_Figure_1.jpeg)

#### Example

Probability of option execution. Due to the shorting time maturity the option becomes increasingly unlikely to be executed. The total probability of execution is 27.7%.

![](_page_53_Figure_4.jpeg)

#### **Parameter Fitting** Introduction Financial Historical timeseries models $\square$ Process $\max_{p} \prod_{i=1}^{M} \prod_{j=1}^{M} (\Theta_{p} A_{i})^{M} \delta(V(s_{t+1}))(V(s_{t}))$ (53)⊞ Portfolio H Measure Sensitivity Implied parameters Distribution Stop time $\min_{p} \sum \left( (\Theta_{p} A_{i})^{M} N(s) - Market Value_{i} \right)^{2}$ (54)Parameter fitting portfolio optimisation Implementation $\square$ $\max_{p} \, (\Theta A_p)^M \, UN$ (55)

![](_page_55_Figure_1.jpeg)

![](_page_56_Figure_1.jpeg)

![](_page_57_Picture_1.jpeg)

# Introduction ⊞ Financial models Implementation $\mathbb{R}^{\mathbb{N}} \to V$ Dadim $\boxplus$ Example

#### What is Dadim

Dadim is a purely discrete computer implementable function class that is isomorph in  $\mathscr{L}^2$  to the function space  $\mathbb{R}^\mathbb{N}\to V$ 

$$\mathcal{K}(V) \cong \mathscr{L}^2(\mathbb{R}^{\mathbb{N}} \to V) \tag{56}$$

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![](_page_59_Picture_4.jpeg)

Dadim operators

Dadim can be defined algebraically as a 5-tuple consisting of a set of Dadims, an underlying vector space and three operators for evaluation, translation and dilatation.

$$\mathsf{K}(V) = (\mathsf{K}, V, \mathsf{T}, \mathsf{T}, \mathsf{Z}) \tag{57}$$

The operators work on the domain as follows. T and Z are vectors of operators whereas each component acts in the according direction.

There are further restrictions on the behaviour of the operators, which are neglected in this talk.

Numerical interface Introduction The domain and the three operators can implemented in **⊞** Financial models Java an interface with three methods. Implementation Dadim // Licensed under the Open Software License v1.1  $\boxplus$  Example public interface Dadim { public Object da(); public void trans(int dir, int length); public Dadim zoom(int dir); }

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#### What is Dadim

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$$\mathcal{K}(V) \cong \mathscr{L}^2(\mathbb{R}^{\mathbb{N}} \to V) \tag{59}$$

![](_page_62_Picture_1.jpeg)

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<ul> <li>Implementation</li> <li>Dadim</li> <li>Example</li> <li>Price</li> <li>Distribution</li> <li>Variance</li> <li>Delta-Hedge</li> <li>Trading costs</li> <li>Best hedge</li> </ul>	Process options: $(S_0, c_0) = (0, 0)$ $\sigma = 0.5$ $\mu = 0$ $r = 0$	Portfolio options: X = 0 M = 4 k = 0.4	(60)
	Consider a european call option with $\pi(S) : \mathfrak{K}(\mathbb{R})$ $\pi(S) = max(0, S)$	n payment $\pi$ : S - X)	(61)
	and a simple process for $S$ and $c$ : $\Theta : \mathcal{K}(\mathbb{R}) \to \mathcal{I}$ $\Theta = T_c^{rc} T_S^{\mu} B_S^{\sigma^2}$	大(ℝ) 2	(62)

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#### Computing the expected value

The expectation value is computed as  $\Theta^M(c + \pi(S))$ :

 $V \in \mathcal{K}(\mathbb{R})$ 

$$V = c + \pi(S)$$
  
for t = 0 to M-1  
$$V = \Theta V$$

$$V = Z^3 V$$
  
Result:

大 V 
$$= 0.389$$

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#### Distribution

$$f(c) = \Theta^M \mathbf{1}_{c > \pi(S)} \tag{63}$$

V = heaviside(c - 
$$\pi(S)$$
)  
for t = 0 to M-1  
V =  $\Theta$  V

$$\mathbf{V} = \frac{d}{dc} \quad \mathbf{V}$$

$$V = Z_c^3 Z_S^3 V$$
  
Result:

![](_page_66_Figure_1.jpeg)

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#### Variance

Let's sell the option at the expected price 0.398 and compute the expected square deviation:

$$\Theta^M \Big( 0.398 + c - \pi(S) \Big)^2 \tag{65}$$

 $var \in \mathcal{K}(\mathbb{R})$ 

var = pow(0.398 + c -  $\pi(S)$ , 2) for t = 0 to M-1 var =  $\Theta$  var

var =  $Z^3$  varResult: $\bigstar$  varsqrt( $\bigstar$  var)= 0.34

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#### Delta-Hedge

The delta-hedge investment is defined by transaction of cach from account c into stocks in deposit h:

$$A_H V = (T_c^{-S} T_h)^{-\frac{d}{dS}V_1}$$
(66)

A two dimensional measure for expected value and variance:

$$(A_H \Theta)^M \begin{pmatrix} 0.398 + hS + c - \pi(S) \\ (0.398 + hS + c - \pi(S))^2 \end{pmatrix}$$
(67)

Although, the delta-hedge is designed for continuous trading, there should be at least some reduction in the variance.

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#### **Delta-Hedge** $V \in \mathcal{K}(\mathbb{R}^2)$ $a = 0.398 + c + h*S - \pi(S)$ V = [a, pow(a, 2)]for t = 0 to M-1 $V = \Theta V$ $\Delta = \frac{d}{dS} V[0]$ $V = T_h^{-\Delta} T_c^{S\Delta} V$ $V = Z_h^2 Z_S^2 V$ Result: 大 V[0] 0.00大 V[1] 0.03sqrt(大 V[1]) 0.17

![](_page_70_Figure_1.jpeg)

#### Transaction costs

This model for the transaction costs assumes, that the price depends linearly on demand. The total costs are therefore proportional to the squared trade size.

![](_page_70_Figure_4.jpeg)

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#### **Transaction costs**

The hedge activity including transaction costs

$$A_H V = (T_c^{-S} T_h)^{-\frac{d}{dS}V_1} \underbrace{T_c^{-k\left(\frac{d}{dS}V_1\right)^2}}_{C} V$$

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## Transaction costs

$$V \in \mathcal{K}(\mathbb{R}^{2})$$
  
a = 0.398 + c + h\*S -  $\pi(S)$   
V = [a, pow(a,2)]  
for t = 0 to M-1  
V =  $\Theta$  V  
 $\Delta = \frac{d}{dS}$  V[0]  
V =  $T_{h}^{-\Delta}$   $T_{c}^{S\Delta-k\Delta^{2}}$  V  
V =  $Z_{h}^{2}$   $Z_{S}^{2}$  V  
Result:  
 $\pm$  V[0] = -0.27  
 $\pm$  V[1] = 0.12  
sqrt( $\pm$  V[1]) = 0.34

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Trading costs
Best hedge

Best hedge  $a = 0.398 + c + h*S - \pi(S)$ V = [a, pow(a, 2)]for t = 0 to M-1  $V = \Theta V$  $\Delta = h^* - h$  $\mathbf{V} = T_h^{-\Delta - k\Delta^2} T_c^{S\Delta} \mathbf{V}$ opt = solve $\left(\frac{d}{dh^*} V[1] = 0, h^*\right)$  $V = T_{h*}^{opt} V$  $V = Z_{h*}^2 Z_S^2 Z_h V$ Result: 大 V[0] = -0.04大 V[1] = 0.05 sqrt(大 V[1]) 0.23=

## A new financial calculus

Things to do Introduction Financial Ħ models Implementation  $\square$ 1. 大ML, an XML format for standardized model description 2. Monte Carlo engine that reads 大ML expressions 3. Increase and verify accuracy 4. Increase the speed of Dadim. (Adaptivity, Sparse grids, Operator sampling) 5. Stylesheet or XSLT to XHTML/MathML 6. Toolbox for standard models and standard products