An analysis of objective inflation expectations and inflation risk premia

by Sara Cecchetti, Adriana Grasso and Marcello Pericoli
Temi di discussione
(Working Papers)

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Number 1380 - July 2022
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ISSN 1594-7939 (print)
ISSN 2281-3950 (online)

Printed by the Printing and Publishing Division of the Bank of Italy
AN ANALYSIS OF OBJECTIVE INFLATION EXPECTATIONS AND INFLATION RISK PREMIA
by Sara Cecchetti*, Adriana Grasso** and Marcello Pericoli*

Abstract

We study euro-area risk-adjusted expected inflation and the inflation risk premium at different maturities, leveraging inflation swaps, inflation options and survey-based forecasts. We introduce a model that features time-varying long-term average inflation and time-varying inflation volatility and we anchor market-based risk-adjusted measures of expected inflation to survey-based inflation forecasts. The results show that medium-term risk-adjusted expected inflation was close to the ECB's aim from 2010 to mid-2014, has since fallen to a low in March 2020 and has risen significantly since the second half of 2021. The medium-term inflation risk premium was positive until 2014 and turned negative since 2015 despite a sharp rise at the end of 2021. The risk-adjusted probabilities of exceeding the ECB's inflation aim and of seeing deflation over the medium term have been low on average.

JEL Classification: C22, C58, G12, E31, E44.
Keywords: inflation density, inflation risk premium, objective probability.
DOI: 10.32057/0.TD.2022.1380

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* Bank of Italy, Economic Outlook and Monetary Policy.
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1 Introduction

In the aftermath of the sovereign debt crisis, euro area inflation and expected inflation have been drifting downwards, raising concerns among market analysts and policymakers about the anchoring of inflation expectations to the European Central Bank’s (ECB) target – see Corsello et al. (2021) for a brief review. The downward trend intensified, first in 2014 and, subsequently, with the outbreak of the Covid-19 pandemic in 2020. This trend reversed in the second half of 2021 as consumer prices and inflation expectations began to rise. Since expected inflation plays such an important role in monetary policy decisions, a timely and reliable estimate of it is essential to define the monetary policy stance and investors’ decisions on portfolio allocations.

Two sets of variables are typically used to infer the risk-adjusted expected inflation and the inflation risk premium. The first set contains information on inflation implied in instruments traded daily on financial markets, such as bonds, index-linked bonds, inflation swaps and inflation caps and floors, and is used to calculate what is known as breakeven inflation. This measure, being priced in traded assets, refers to a representative risk-neutral investor and is made up of a component defined as risk-adjusted (or objective) expected inflation and a component that rewards the uncertainty borne by the investor, the inflation risk premium.\(^1\) The second set of variables is obtained from analysts’ surveys on expected inflation over different horizons, conducted on a monthly or quarterly basis by specialized agencies or central banks, and provides a risk-adjusted (or objective) measure of expected inflation, or already adjusted for the inflation risk premium. Often the two measures not only differ due to the wedge imposed by the presence of the risk premium, but also go in different directions.\(^2\) Therefore one may wonder which of the two measures gives the correct signal, which is crucial for both investors and monetary authorities.

The aim of this paper is to retrieve reliable estimates of objective expected inflation and the inflation risk premium at high frequency exploiting the information content of both financial market prices and survey data. To the best of our knowledge, this is the first paper that incorporates these features, providing an innovative contribution to literature on euro-area inflation along several dimensions. Our study is of particular interest for both policymakers and investors. The former can estimate the objective inflation expectations, evaluate the degree of anchoring of these expectations to the central bank’s target and possibly assess the sensitivity of medium-term and long-term inflation expectations to shocks affecting short-term

\(^1\)Theoretically, there might be also a liquidity premium component in some measures of breakeven inflation but this component is difficult to quantify and is not the subject of our analysis (nor of most of the literature).

\(^2\)See Cecchetti et al. (2021) for a discussion of the measures of inflation expectations derived from the prices of financial instruments indexed to inflation and from the surveys of professional forecasters.
ones and to inflation surprises (Miccoli and Neri, 2019; Corsello et al., 2021). The latter can use the measure of expected inflation to recover the discount rate in real terms to value fixed-income portfolios over long horizons.

Derivatives-based models of expected inflation extensively used by central banks have important drawbacks. First, they assume that the pricing factors have a constant mean and, therefore, are unable to capture regime changes at low frequencies. Second, they assume a constant volatility of inflation while in reality there is a relationship between price dynamics and their volatility. Empirical evidence shows that, similar to interest rates, low inflation volatility is associated with a low inflation risk premium and with low expected inflation, while high volatility is associated with high levels of inflation. We model expected inflation with the aim of overcoming these drawbacks.

We use daily data of inflation swaps, which provide a direct measure of average breakeven inflation, and inflation caps and floors, which provide information on the entire distribution of risk-neutral expected inflation over different horizons. We complement the information coming from the markets with survey-based measures of inflation expectations, using quarterly data of the survey of professional forecaster (SPF) conducted by the ECB. Assuming the absence of arbitrage, we are able to estimate the parameters driving the dynamics of the risk-adjusted (or objective) expected inflation and its density function. Our estimates cover the period between October 2009 and December 2021.

The literature on expected inflation, risk of deflation, decoupling between expected short-term and long-term inflation, and inflation risk premium is enormous. Typically, researchers have studied the breakeven inflation implied in nominal coupon bonds and in index-linked coupon bonds – see Abrahams et al. (2016), Adrian et al. (2013), Buraschi and Jiltsov (2005), Christensen et al. (2016), Christensen et al. (2012), Wright (2014) for a review – or one measured by inflation swaps – see Haubrich et al. (2012), Camba-Méndez and Werner (2017), Fleckenstein et al. (2017). Cecchetti et al. (2015) use the information content of inflation options to estimate risk-neutral probability densities and investigate signals of decoupling between expected short-term and long-term risk-neutral expected inflation. Regarding the estimate of the inflation risk premium, in the literature there are several models that have given different results in terms of magnitude and even sign – see Adrian et al. (2013), Pericoli (2014), Casiraghi and Miccoli (2019).

Our approach follows the traditional literature on factor models for estimating

---

3See, among others, Adrian et al. (2013) and Joslin et al. (2011).
4See Abrahams et al. (2016) and references therein.
5They find that, since mid-2014, negative tail events affecting short-term inflation expectations have increasingly been channelled towards long-term views, triggering both downward revisions in expectations and upward shifts in uncertainty; on the other hand, the short-term positive tail events mostly left the long-term moments unchanged. This asymmetrical impact may signal a bottom-up disanchoring in long-term inflation expectations.
the term structure of interest rates and assumes that three factors are able to provide a good representation of inflation fluctuations. These factors are "instantaneous inflation", "long-term inflation" and "expected volatility of inflation". Moreover, our factor model exploits the informative content of inflation swaps and inflation options and features time-varying long-term mean and stochastic volatility of inflation. In particular, our paper follows both the Heston (1993) stochastic volatility option pricing model and the long-run risk consumption model of Bansal and Yaron (2004). Joining these models, we obtain as in Fleckenstein et al. (2017), a framework where inflation can have fluctuating uncertainty and a small, predictable long-term component. In general, this allows for a wide range of possible time-series properties for realized inflation and the inflation risk premium. Our main innovation compared to the work of Fleckenstein et al. (2017) consists of anchoring the risk-adjusted (or objective) expected inflation implied in the model to analysts’ surveys in order to link it with observed risk-adjusted expected inflation, in the spirit of Joyce et al. (2010) and Kim and Orphanides (2012). Therefore, our model distinguishes itself from those proposed in the literature for the euro area – see, among others, Camba-Méndez and Werner (2017), Casiraghi and Miccoli (2019) and Pericoli (2014). Furthermore, the assumption of the absence of arbitrage allows to calculate not only the risk-neutral but also the risk-adjusted (or objective) probability density function of expected inflation over different horizons, and the probability that inflation is below or above a certain threshold.

In a nutshell, our results show that the introduction of a time-varying long-term average inflation (the value to which inflation should revert over the long term, presumably influenced by the ECB target) and variable volatility of inflation makes it possible to capture regime changes by improving the estimates of long-term expected inflation. Risk-adjusted long-term expected inflation was close to the ECB’s target from 2010 to mid-2014 but declined thereafter, with only temporary increases favored by new waves of unconventional monetary policies, reaching a low in March 2020. Since the second half of 2021, it has increased significantly to reach 2.2% by the end of 2021. The inflation risk premium was positive in the first part

---

6In a continuous time model, this is the annualized continuously compounded inflation rate that we expect to be realized for an infinitesimally short period of time, one day in our analysis.

7We impose that the estimates are close to the inflation expected by analysts surveyed by the central bank, for the available maturities.

8The risk-neutral probability is the probability of potential future outcomes that contains the inflation risk premium. The term risk-neutral can sometimes be misleading because some people may assume it means that the investors are neutral, unconcerned, or unaware of risk. In contrast, the risk-neutral probability accounts for the investors’ aversion to risk: in general, the risk-neutral probability tends to assign more weight to outcomes investors are worried about, attributing higher probability to extreme events such as deflation or very high inflation. Mathematically, the risk-neutral probability is the implied probability measure derived from the observable prices of the relevant instruments, defined using a risk-neutral utility function and assuming absence of arbitrage. On the other hand, the risk-adjusted (or objective) probability can be inferred from historical data, being estimated from the past dynamics of prices and other financial variables.
of the sample but turned negative in 2014, hitting a low after the outbreak of the pandemic in 2020. It returned to zero in the autumn of 2021. The probability of inflation being negative over a 3-year horizon peaked above 50% in late 2014 and in early 2020, with the outbreak of the pandemic. Conversely, the likelihood that inflation could exceed the ECB’s 2% target was always below 40%, with the exception of the 2011-12 period and the second half of 2021 when it rose to more than 80%.

The anchoring of objective inflation expectations to those of analyst surveys is of fundamental importance in the calibration of the model. The omission of this anchor gives inflation expectations which rise too sharply in the latter part of the estimated period, consequently causing an unjustified decrease in the inflation risk premium.

The paper is structured as follows. Section (2) presents the model, the identification strategy and the data. Section (3) documents the results of our estimates while Section (4) compares them with those already in the literature. Section (5) concludes.

2 The model

We follow Bansal and Yaron (2004) and Fleckenstein et al. (2017) and estimate the following model

\[
\begin{align*}
    i(X, Y, T) &= A(T) + B(T) \cdot X + C(T) \cdot Y + \varepsilon(T), \\
    m(V, T) &= G(T) + H(T) + U(T) \cdot V + \varepsilon(T),
\end{align*}
\]

where \(i(X, Y, T)\) is the inflation swap rate with maturity \(T\);\(^9\) also referred to as \(i_T\) below, \(m\) is the volatility estimated from the risk-neutral density implied in options with expiration \(T\). \(X, Y, V\) are three factors that drive inflation swaps and the option volatility. \(A(T), B(T), C(T), G(T), H(T), U(T)\) are vectors to be estimated. In keeping with the spirit of term structure models, these vectors have a recursive structure – see appendix (A.3 and A.4). We consider 15 maturities for inflation swaps, i.e. 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 12, 15, 20, 25, 30, and 10 maturities for options volatility, i.e. 1, 2, 3, 5, 7, 10, 12, 15, 20, 30. The model we estimate, that is the formula for the swap price and the volatility of the options, is obviously determined by the dynamics of the factors and we report in the appendix both the derivation of the swap price and the implied volatility of the option, and in particular why we can separate \(G(T)\) and \(H(T) + U(T) \cdot V\) in equation (2). The dynamics of inflation \(dI\), where \(I\) is the consumer price index, and the three factors under the

\(^9\)The inflation swap rate \(i(X, Y, T)\) is equal to the expected inflation rate over the horizon \(T\) under the risk neutral measure \(Q\), i.e. \(i(X, Y, T) = \mathbb{E}^Q[I_T/I_0 - 1]\) where \(I_k\) is the consumer price index at time \(k\). It relates to the swap price \(F(X, Y, T)\) with the formula \(F_T = (1 + i_T)^T\).
risk-adjusted (or objective) probability measure $\mathbb{P}$ is\(^{10}\)

\[
\begin{align*}
    dI &= X \cdot Idt + \sqrt{V} I \cdot dZ^P_I \\
    dX &= \kappa(Y - X)dt + \eta dZ^P_X \\
    dY &= (\mu - \xi Y)dt + sdZ^P_Y \\
    dV &= (\delta - \psi V)dt + \sigma \sqrt{V} dZ^P_V
\end{align*}
\]

where $Z^P_I, Z^P_X, Z^P_Y, Z^P_V$ are uncorrelated Brownian motions. The model (3-6) incorporates the factor $X$,\(^{11}\) which represents instantaneous expected inflation, that in the long run will evolve around the level of factor $Y$; the latter then represents the long-run trend (or long run mean) of inflation. Furthermore, inflation $dI$ is determined non only by $X$ and indirectly by $Y$ but also by $V$, a variance factor that follows a stochastic process. The setting is reminiscent of Fleckenstein et al. (2017) and Bansal and Yaron (2004) setup for instantaneous and long-run consumption and the Heston (1993) model for interest rates with stochastic volatility. In particular, the volatility of inflation has two components: the volatility due to the variation in expected inflation $X_t$ and the volatility resulting from unexpected inflation, driven by the state variable $V_t$. The model (3-6) has a counterpart under the risk-neutral $\mathbb{Q}$ probability measure, i.e.:

\[
\begin{align*}
    dI &= X \cdot Idt + \sqrt{V} I dZ^Q_I \\
    dX &= \lambda(Y - X)dt + \eta dZ^Q_X \\
    dY &= (\alpha - \beta Y)dt + sdZ^Q_Y \\
    dV &= (\theta - \phi V)dt + \sigma \sqrt{V} dZ^Q_V
\end{align*}
\]

The connection between (7-10) and (3-6) is guaranteed by the assumption of a system of market prices of risk\(^{12}\) that allows to obtain for each state variable the same dynamics under both probability measures, imposing standard conditions to exclude arbitrage opportunities. Note that even if the functional form of the drift for the $I$ process is the same under the objective – equation (3) – and risk-neutral measures – equation (7) – this does not imply that the expected value is the same under $\mathbb{P}$ and $\mathbb{Q}$, because it is related to the different dynamics of $X, Y \text{ and } V$.

---

\(^{10}\) $\mathbb{P}$ is the probability measure not adjusted for the risk perceived by investors that can be inferred by historical data.

\(^{11}\) The stochastic process followed by this factor is an analogue of the stochastic differential equation with which the instantaneous interest rate evolves in Vasicek’s model.

\(^{12}\) See Appendix A.1 for the derivation of the system of market prices of risk and the relationships between the drift parameters under the two probability measures.
The model (7-10) states that inflation can be written as

\[
I_T/I_0 = \exp(w_T + u_T)
\]

\[
w_T = \int_0^T X_t dt
\]

\[
u_T = -\frac{1}{2} \int_0^T V_t dt + \int_0^T \sqrt{V_t} dZ_t
\]

where the mean and the variance of \( w_T \) and \( u_T \) are shown in the appendix A.4.

We use the Heston (1993) model to derive the density of \( u_T \) that we use to price options. We also estimate the parameters of the processes (3-6) under the objective probability measure to recover the probabilities that expected inflation is above or below a certain threshold.

### 2.1 Identification

The factors are obtained using a completely standard approach, which is entirely based on the existing methodology widely applied in the literature. Following Chen and Scott (1993), Duffe and Singleton (1997) and Duffee (2002) we solve for the values of \( X, Y, \) and \( V \) from specific inflation swaps and option volatilities, and then jointly estimate the parameters of both the risk-neutral and objective dynamics for these variables using maximum likelihood. The unobservable factors are extracted by inverting the measurement equation by assuming that a number of assets equal to the number of factors is observed without error. In particular, we assume that the two inflation swaps with a maturity of 2 and 30 years and the implied volatility in the prices of the 3-year options are priced without errors,\(^\text{13}\) namely \( \varepsilon_2 = 0, \varepsilon_{30} = 0, \varepsilon_3 = 0. \) Therefore, the three factors, \( X, Y, V, \) are obtained by inverting the linear system (1-2) for \( i = 2, 30 \) and \( j = 3. \) The other inflation swaps and volatilities are priced with errors.

Furthermore, in order to estimate a reasonable value of expected inflation, we anchor the objective expected inflation to analysts surveys; specifically, we assume that the 1-year forward inflation rate in 0, 1 and 4 years time\(^\text{14}\) implied in the inflation swaps estimated under the \( P \) measure, \( f^P(X, Y, l) \) with \( l = 0, 1, 4, \) are close, up to a measurement error, to the 1-year inflation expected by analysts surveyed by the central bank, \( E^{SPF}(I_{l+1}/I_l), \) i.e.

\[
f^P(X, Y, l) = E^{SPF}(I_{l+1}/I_l) + z_l, \quad \text{for } l = 0, 1, 4. \quad (12)
\]

We assume that the errors have distribution \( \varepsilon \sim N(0, \Sigma), \varepsilon \sim N(0, \Psi) \) and \( z \sim N(0, \Omega) \) and that the covariance matrices are diagonal, i.e. the errors are

\(^\text{13}\)For the choice of the assets assumed perfectly priced we follow Fleckenstein et al. (2017).
\(^\text{14}\)These are the maturities for which survey data are available.
uncorrelated, and denote the main diagonal elements with \(\Sigma_i, \Psi_j, \Omega_l\). Then, we maximize the sum over all values of \(t\) of the log-likelihood function (13) conditional on the data over the 38-dimensional parameter vector \(\Theta = \{\alpha, \beta, \lambda, \kappa, \mu, \xi, \sigma, s, \eta, \phi, \psi, \delta, \theta, \Sigma_1, \Sigma_3, \Sigma_4, \Sigma_5, \Sigma_6, \Sigma_7, \Sigma_8, \Sigma_9, \Sigma_{10}, \Sigma_{12}, \Sigma_{15}, \Sigma_{20}, \Sigma_{25}, \Psi_1, \Psi_2, \Psi_5, \Psi_7, \Psi_{10}, \Psi_{12}, \Psi_{15}, \Psi_{20}, \Psi_{30}, \Omega_1, \Omega_2, \Omega_3\}\) using a standard simplex algorithm.

\[
L_t = -(25/2) \ln(2\pi) + \ln(k) + \ln|J_{t+\Delta t}| - \frac{1}{2} \ln |\Sigma| - \frac{1}{2} \epsilon_t^t \Sigma^{-1} \epsilon_t \Delta t \tag{13}
\]

\[
= -\frac{1}{2} \ln |\Psi| - \frac{1}{2} \epsilon_t^t \Psi^{-1} \epsilon_t \Delta t - \frac{1}{2} \ln |\Omega| - \frac{1}{2} \zeta_t^t \Omega^{-1} \zeta_t \Delta t
\]

\[
- \ln \left( 2\pi \sigma_X \sigma_Y \sqrt{1 - \rho^2_{XY}} \right) - \frac{1}{2} \left( 1 - \rho^2_{XY} \right) \left[ \left( \frac{X_t + \Delta t - \mu_{X_t}}{\sigma_X} \right)^2 \right]
\]

\[
-2\rho_{XY} \left( \frac{X_t + \Delta t - \mu_{X_t}}{\sigma_X} \right) \left( \frac{Y_t + \Delta t - \mu_{Y_t}}{\sigma_Y} \right) + \left( \frac{Y_t + \Delta t - \mu_{Y_t}}{\sigma_Y} \right)^2 \right]
\]

\[
- k(V_t + \Delta t + V_t e^{-\psi \Delta t}) + \frac{1}{2} q(\ln V_t + \Delta t - \ln V_t + \psi \Delta t)
\]

\[
+ \ln I_q \left( 2k \sqrt{V_t + \Delta t} V_t e^{-\psi \Delta t} \right)
\]

where \(I_q(.)\) is the modified Bessel function of the first kind of order \(q = 2\delta/\sigma^2 - 1\) and \(k = 2\psi/(\sigma^2(1 - e^{-\psi \Delta t}))\). \(J\) is the Jacobian of the linear mapping from the two inflation swaps and the implied volatility into \(X, Y\) and \(V\). Note that, as usual in this literature, the errors terms \(\varepsilon, e\) and \(z\) in equation (13) are valued under the \(Q\) measure while the three factors, \(X, Y, V\), under the \(P\) measure.

This methodology makes it possible to jointly estimate the parameters of both the risk-neutral and objective dynamics and to recover the price of the risk inherent in the prices of inflation securities. We test for the existence of risk premia by examining whether the five parameters that appear in the objective dynamics of \(dI_t\) are equal to the corresponding parameters in the risk-neutral dynamics.

### 2.2 Option prices

We do not directly use caps and floors quotes for the estimation of model (1, 2, 12), while we use the standard deviation calculated from option price risk-neutral density function as in Cecchetti et al. (2015). Once the parameters have been estimated, we can retrieve the probability density function also under the objective measure \(P\), and consequently the corresponding objective prices.

Cap, \(C\), and floor, \(P\), prices are defined by

\[
C(X, Y, V, T; K) = D(T) \cdot E^Q[\max(0, (1 +i_T)^T - (1 + K)^T)]
\]

\[
P(X, Y, V, T; K) = D(T) \cdot E^Q[\max(0, (1 + K)^T - (1 + i_T)^T)]
\]
where $D(T)$ is the discount factor and $Q^*$ is a forward measure. Under $Q^*$, $w_T$ is normally distributed with mean $\mu_u = \ln((1 + i_T)^T) - \frac{1}{2}\sigma_w^2$ and variance $\sigma_w^2$, and $u_T$ has a known distribution function $h(u_T)$ that we recover as a special case of the Heston (1993) model, with mean $\mu_u$ and variance $\sigma_u^2$. The discount factor $D(T)$ is defined as

$$D(T) = E^Q\left[e^{-\int_0^T r_s ds}\right] \quad (14)$$

where $r_t$ is the nominal instantaneous riskless interest rate, which is the sum of the real riskless interest rate $R_t$ and instantaneous expected inflation $X_t$:

$$r_t = R_t + X_t. \quad (15)$$

Assuming that $R_t$ and $X_t$ are uncorrelated and that $R_t = 0$, cap and floor prices can be written as (see appendix A.6)

$$C = \frac{1}{(1 + i_T)^T} \int_{-\infty}^{+\infty} \left[(1 + i_T)^T N(a_1)e^{u_T} - (1 + K)^T N(a_2)\right] h(u_T) du_T \quad (16)$$

$$P = \frac{1}{(1 + i_T)^T} \int_{-\infty}^{+\infty} \left[(1 + K)^T N(-a_2) - (1 + i_T)^T N(-a_1)e^{u_T}\right] h(u_T) du_T \quad (17)$$

where

$$a_1 = \frac{w_T - T \ln(1 + K) + T \ln(1 + i_T) + \frac{1}{2}\sigma_w^2}{\sqrt{\sigma_w^2}}$$

$$a_2 = a_1 - \sqrt{\sigma_w^2}$$

$K$ is the strike price and $N(\cdot)$ is the normal cumulative distribution function.

We confirm the correctness of the pricing formula for swaps, caps and floors in Figures (1), (2) and (3) showing the convergence of a Monte Carlo simulation with 3,000 extractions. The results show that the simulation replicates after less than 1,000 steps the prices obtained by the closed formulas (1), (16) and (17)\(^{16}\).

We also verify that the prices of the options implied in the model are close to the observed prices of the options.

Furthermore, we use the Gram-Charlier expansion to approximate the distribution function of the logarithm of inflation $\ln I_T = w_T + u_T$; we will use this approximation to estimate the density function of inflation under the objective probability measure.\(^{17}\) Define $x$ the standardized value of inflation, with probability density

---

\(^{15}\)The forward measure $Q^*$ is defined by the dynamics of $I$ given by equations (7), (8), (9) and (10), where the drift in (8) is augmented by $\eta^2B(\tau)dt$ and the drift in (9) is augmented by $s^2C(\tau)dt$, with $\tau = T - t$. See Brigo and Mercurio (2006) for a detailed explanation of the convenient definition of the forward measure.

\(^{16}\)For the swap, the Monte Carlo exercise shows the convergence of the swap price $F_T = (1 + i_T)^T$.

\(^{17}\)This approximation, also used by Fleckenstein et al. (2017), avoids the need, costly from a computational point of view, to calculate all the prices of the options under the objective probability measure and estimate the objective densities from these risk-adjusted prices.
function $f(x)$ and the first two cumulants $c_1 = \mu_w + \mu_u$ and $c_2 = \sigma_w^2 + \sigma_u^2$. The density function $f(x)$ can be approximated by

$$f(x) \approx \left[ 1 + c_1 x + \frac{1}{2} (c_1^2 + c_2 - 1) (x^2 - 1) \right] \cdot n(x)$$

where $n(\cdot)$ is the normal probability density function and $[1, x, x^2-1]$ are the Hermite polynomials up to the second order. $c_1$ and $c_2$ are functions of the parameters of model (1, 2, 12) and are shown in the appendix A.5.

### 2.3 Data

We use daily data from October 2009 to December 2021 for inflation swaps with maturities 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 12, 15, 20, 25, and 30 years.

Instead of using the full spectrum of option quotes with the corresponding maturity and strike prices, we summarize for each set of options with the same maturity their volatility by calculating the standard deviation of the inflation risk-neutral density function obtained from the option prices. This methodology makes it possible to reduce the number of parameters and to use a single time series of volatility per maturity, ignoring the presence of possible non-linearity in prices. We estimate the daily density function using zero caps and zero floors on euro-area HICP with maturities 1, 2, 3, 5, 7, 10, 12, 15, 20, 30 years and 17 strikes that range from -3% to 6%, for a total of around 170 time series. The density functions are estimated with the Cecchetti et al. (2015) and Taboga (2016) methodology\(^\text{18}\) and the standard deviation of the density is obtained by numerical integration.

We can say that inflation swaps and options that we use can be considered fairly liquid, as discussed in Fleckenstein et al. (2017),\(^\text{19}\) so liquidity issues should not affect our estimates.

The analysts’ forecasts are derived from the quarterly survey of professional forecasters (SPF) conducted by the ECB, which are transformed into daily data assuming that the forecasts are constant until the new release.\(^\text{20}\) From SPF we use 1-year expected inflation, 1-year forward expected inflation after one year and the mean of the aggregate probability distribution of 1-year forward expected inflation after 4 years.

\(^{18}\)This methodology, as shown in Taboga (2016), is robust even in periods of low liquidity, e.g. during market turmoil.

\(^{19}\)See Section 2 in their paper.

\(^{20}\)We also tried other smoother forms of interpolation to convert quarterly data into daily data, but results where broadly unchanged.
3 Results

3.1 Parameters and fitting

The estimates of the parameters of model (1, 2, 12) and the standard deviations – calculated with the Huber sandwich estimator\(^{21}\) – are presented in Table (1). The 38 parameters are highly significant with low p-values, except for the parameter \(\theta\),\(^{22}\) related to the long-term mean \((\frac{\theta}{2})\) of the volatility under the \(\mathbb{P}\)-measure. Notably, parameters under the \(\mathbb{Q}\) measure and the \(\mathbb{P}\) measure are significantly different. In particular, as regards to the factor \(V\), the speed of mean reversion \(\psi\) is much larger under the \(\mathbb{P}\) measure that the corresponding parameter \(\phi\) under the \(\mathbb{Q}\) measure, resulting in a volatility under the \(\mathbb{P}\) measure lower than that under the \(\mathbb{Q}\) measure. The latter result is consistent with what is expected from a theoretical point of view as the volatility under the risk-neutral measure reflects also the variability of the risk premium component. Moreover, looking at the process of the factor \(Y\) corresponding to the long-run trend of inflation, the long-term mean under the \(\mathbb{P}\) measure \((\frac{\theta}{2})\) is 1.98%, while the corresponding long-term mean under the \(\mathbb{Q}\) measure \((\frac{\theta}{2})\) is 2.98%. This result is in line with what we expected, as the average positive slope of the inflation swap curve observed in the market during the review period corresponds to a positive risk premium; the latter translates into a value of the risk-neutral long-term mean higher than the objective one.

Table (2) reports the pricing errors and the root mean squared errors for the inflation swaps and Table (3) the counterparties for the implied volatilities. Overall, the estimates of model (1, 2, 12) give a good approximation of the prices for both inflation swaps (with a pricing error in the range \([-5.15, +7.50]\) basis points) and implied volatility (with a pricing error in the range \([-5.42, +4.75]\) basis points). The goodness of fit can be valued by Figure 4 for inflation swaps and Figure 5 for implied volatility. The fitting for the former is particularly good, while for the latter we observe higher errors due to the fact that one factor may not be sufficient to cross-sectionally fit the term structure of implied volatilities.

In the spirit of the standard affine term structure models, we present the vectors \(A\), \(B\) and \(C\), which appear to have the usual form interpretable as level, slope and curvature and the vectors \(G\), \(H\) and \(U\), which are the corresponding loadings for the factor \(V\) (Figure 6).

3.2 Estimated factors and implied volatility

We report the estimates of the factors \(X\) and \(Y\). The instantaneous expected inflation \(X\) shows large fluctuations from 2010 to the end of 2021 with a first peak in 2011 at

\(^{21}\) We have chosen the Huber sandwich estimator (see Huber (1967)) because it is a robust estimator. However, we also calculated other statistical tests, that gave similar results.

\(^{22}\) This parameter is however still significant at a 90 per cent confidence level.
2.4%, a decline to -0.8% in 2015 and to -0.7% in 2020, and a maximum in December 2021 at 3.1% (Figure 7). Longer-term inflation \( Y \), on the other hand, remained more stable around 2% from 2010 to 2015, falling to 0.8% in 2016, rising around 1.5% in the 2017-2019 period, decreasing to below 1% from mid 2019 to early 2021, and finally rising to 2% in autumn 2021.

We also report the estimates of the implied volatility \( m \) in equation (2) under the measure \( Q \) and \( P \) (Figure 8). For all maturities it is obtained that those under the \( P \) measure are lower than those under the \( Q \) measure, which means that objective inflation tends to be less volatile than breakeven inflation.

### 3.3 Expected inflation and inflation risk premium

We use the model (1, 2, 12) to calculate inflation swaps under the \( P \) measure – also labeled as expected inflation – and the inflation risk premium. The results for the 1-year, 3-year, 5-year, 10-year and 5-year five year forward maturity are presented in Figure 9. Let us take the 3 and 5-year maturity as representative of the medium-term outlook. For the 3-year maturity, the inflation risk premium, i.e. the difference between the market value of the inflation swap and the model value of the inflation swap estimated under the \( P \) measure, averages around 20 basis points from 2010 to 2015, drops to zero from 2015 to 2016, becomes negative between 2016 and 2017, goes back to zero until 2019, becomes negative again until the last quarter of 2021, and finally rises to approach zero. For the 5-year maturity, the dynamics of the inflation risk premium is similar but the size is larger in absolute values.

The results show that the introduction of a variable long-term inflation improves our understanding of expected inflation as we are able to achieve large variability in expected inflation and inflation risk premium at the same time.\(^{23}\) Expected inflation over the long term (10-year horizon) was close to the ECB’s target from 2010 to mid-2014 but has subsequently declined reaching a minimum in March 2020, with temporary increases as the ECB adopted quantitative measures to avoid the materialization of a deflationary scenario.\(^{24}\) In particular, at the beginning of 2015 with the launch of the Asset Purchase Programme (APP), in the first quarter of 2016 with the increase in the pace of monthly purchases of government bonds under the APP from 60 to 80 billion euros, and early 2020 with the launch of the Pandemic Emergency Purchase Programme (PEPP). Thereafter, the declining trend reversed and long-term objective expected inflation markedly increased in particular in the second half of 2021, reaching 2.2% at the end of the year.

Conventional asset pricing theory suggests that the sign of an asset’s risk premium depends on the sign of the covariance of its return with the consumption or

\(^{23}\)This peculiarity is not obtained with other commonly used models; see section 4 for a comparison with other results available in the literature.

\(^{24}\)See Schnable (2021) for estimates of survey- and market-based measures of inflation expectations and a discussion of their dynamics in recent years.
wealth of the representative investor. Accordingly, assets with payoffs tied to inflation have a risk premium linked to the covariance of inflation with output growth. Therefore, the inflation risk premium turns out to be positive when high inflation is associated with poor economic outcomes – i.e. the covariance between inflation and growth is negative. This result is consistent with a predominance of economic shocks that move inflation and real growth in opposite direction, such as ‘supply-side’ shocks – like an oil price increase – that simultaneously raise inflation and lower real consumption. Conversely, a negative inflation risk premium is consistent with an increasing role for ‘demand-side’ shocks that instead push inflation and real economic activity in the same direction, such as that observed in the euro area since 2015.

Our findings confirm our theoretical assumptions. The inflation risk premium, positive until 2014, became negative after 2015 and reached a minimum after the outbreak of the pandemic in early 2020, to go back to values close to zero in autumn 2021. Bulligan et al. (2021) discuss the comovement between the sign of the inflation risk premium and the correlation between inflation expectation and expected growth.

The inflation risk premium increases with maturity in absolute values but on average shows negative values across the maturity spectrum between 2015 and 2017 and between early 2019 and late 2021. Figure 10 shows the average inflation risk premium and the 5th and 95th percentile in relation to maturity. Considering the 5-year maturity as representative, compared to an average value of approximately −2.5 basis points, the 5th percentile of the inflation risk premium is equal to −36 basis points and the 95th percentile to 26 basis points. Overall, the average value of the inflation risk premium between 2009 and 2021 is increasing in relation to maturities since the 10-year maturity but remains at modest levels, ranging between −4 and 12 basis points.

The anchoring of objective inflation expectations to those of analysts’ surveys is of paramount importance in model calibration. As a robustness exercise we have also tried to estimate the model without this anchoring, obtaining inflation expectations that in the last part of the estimated period rise too sharply, consequently resulting in an unjustified decrease in the inflation risk premium.

### 3.4 Probability density functions and probability of deflation/inflation

The model allows to compute the probability density function under the measure $\mathbb{P}$ using the approximation presented in equation (18). Figure 12 shows the densities for the 2, 3, 5, and 10 year maturities. These densities allow us to calculate the probability that inflation falls within selected ranges over a defined horizon under measure $\mathbb{P}$. Since the risk of observing deflation or exceeding the ECB target are among the most debated topics among researchers and policy-makers, we report, on the one hand, the probability that inflation is negative and, on the other hand,
the probability that it exceeds 2% over a mid-term horizon under both measures $P$ and $Q$ in Figure 13. We consider the three-year horizon as representative for medium-term inflation expectations.

The probability under measure $P$ that inflation is negative in three years is negligible from 2009 to mid-2014 and from 2017 to early 2020. It exceeds 50% in late 2014 and in early 2020 with the outbreak of the pandemic. Since 2009, the probability under measure $P$ is lower than that under measure $Q$ during quiet periods, while it is higher during times of crisis. One possible explanation for this empirical evidence is that investors are less concerned about the risk of deflation than that of very high inflation, and thus underestimate the actual probability of deflation. The objective probability of inflation exceeding the ECB target of 2% is less than 40%, except for the 2011-12 period when it reaches 70%, until September 2021; in the last quarter of 2021 it rose to very high values, exceeding 80% per cent in December. In general, the probability of inflation exceeding 2% is similar in the two measures but looking at the peaks observed at the beginning and at the end of the review period, while in the first the probability under measure $P$ was lower than that under measure $Q$, in the most recent peak the probability under measure $P$ was higher than that under measure $Q$. Investors would therefore have perceived the risk of high inflation during the sovereign debt crisis as more worrying than in the recent period.

4 Comparison

4.1 Survey forecasts

We compare our market-based risk-adjusted expected inflation with the risk-adjusted measure surveyed by the ECB SPF for the 1-year forward maturity in 1-year time and in 4-year time$^{25}$ (Figure 11). In general, SPF survey-based risk-adjusted expected inflation is higher than breakeven inflation suggesting that the inflation risk premium is on average negative and that, according to this measure, investors are willing to pay an insurance against low or negative inflation.

We find that the survey-based measure is higher than market-based until the end of 2020 such that the survey-based inflation risk premium is not only negative but also lower than the market-based inflation risk premium. However, during 2021 the survey-based expected inflation became lower than the market-based measure.

In particular, the 1-year 1-year forward SPF survey-based expected inflation closely follows the corresponding breakeven inflation until 2012 producing a small and stable inflation risk premium, which turns negative and large in absolute terms between 2015 and 2020 (left panel). From 2020, the 1-year 1-year forward SPF survey-based measure is lower than the market-based one. Instead, 1-year 4-year

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$^{25}$They represent one-year inflation expected in one year and four years, respectively.
forward SPF survey-based risk-adjusted expected inflation is higher than market-based measure from 2012 to 2018 but is generally closer to the market-based estimate (right panel). The difference between the two risk-adjusted estimates switches sign from 2020, becoming positive.

In general, the difference between the two measures widens when inflation decreases – as between end 2012 and 2016, and between end 2018 and the first quarter of 2020 – while it decreases when inflation rises. At the end of 2021, after a marked rise in inflation observed in 2020 and 2021, the difference between the market-based estimate (2.07%) and the survey-based estimate (1.85%) peaked to 0.22 percentage points on December 2021.

### 4.2 Other models

We compare our results with those obtained from the models prevalent in the literature and extensively used at central banks. Figure 14 presents the expected inflation and the inflation risk premium for the 5-year 5-year ahead maturity estimated with the methodology of this paper (CGP), with those obtained by replicating the methodology of Adrian et al. (2013) (ACM), Joslin et al. (2011) (JSZ) and Pericoli (2014) (PER). The main difference between the CGP and the ACM and JSZ models can be ascribed to the greater sensibility of the CGP’s model to potential regime changes. This is attained especially by allowing for stochastic volatility and variable long-term average inflation while also disciplining long-term objective inflation expectations via SPF survey data; the long-term average expected inflation estimated with the ACM and JSZ models is instead constrained and set at 1.9%, close to the ECB’s long-term inflation target; moreover, CGP estimates the model up to the last date available (while in the other two models parameters are estimated up to April 2021). Therefore, the estimates of the objective component of expectations over the entire estimation period tend to vary more over time in the CGP’s model compared to the ACM and JSZ models. CGP expected inflation is similar to that estimated with the PER methodology and extremely different from ACM expected inflation, which is surprisingly stable. JSZ expected inflation, a measure adopted by the ECB staff as published in Camba-Méndez and Werner (2017), is less variable and very close to CGP and PER estimates. As for the inflation risk premium, the CGP premium has a similar dynamics to the ACM one, while is different from the PER premium, which is smoother and lower in absolute value when the other models move further away from zero. The JSZ risk premium is very similar to that estimated by CGP. Overall, the CGP estimates seem to provide a reasonable description of expected inflation thanks to the introduction of long-term variable average inflation instead of setting long-term expected inflation at a particular value.

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26 The PER model is the only that uses nominal and index-linked bonds to breakdown risk-neutral expected inflation.
5 Conclusion

We propose a factor model to measure expected inflation and the inflation risk premium that leverages two sets of market instruments, inflation swaps and inflation options, and that takes into account analysts’ survey forecasts of inflation to anchor risk-adjusted (or objective) measures of inflation. The model specification, featuring variable long-term average inflation and inflation volatility, allows for a fairly general structure of inflation expectations and makes it possible to capture changes in regime. Moreover, our framework makes it possible to compute not only the risk-neutral but also the risk-adjusted probability density function of expected inflation over different horizons.

The results show that long-term expected inflation were close to the ECB’s 2% target from 2010 to mid-2014, but has since fallen reaching a low in March 2020 with only temporary increases favored by new waves of unconventional monetary policies. The decline in inflation expectations stopped with the reopening of economies after the most acute phase of the pandemic when supply bottlenecks, increases in commodity prices and the revision of the ECB monetary policy strategy led to a marked increase in inflation. In particular in the second half of 2021 risk-adjusted long-term expected inflation increased significantly, rising by just over 2%. The inflation risk premium, positive until 2014, has turned negative since 2015 reaching a low after the outbreak of the pandemic in early 2020, to return to values close to zero in autumn 2021. The risk-adjusted probability of inflation being negative over a 3-year horizon peaked above 50% in late 2014 and early 2020 with the outbreak of the pandemic. Conversely, the risk-adjusted probability that inflation could exceed the ECB’s 2% target over a 3-year horizon was always below 40%, except for the 2011-12 period and the second half of 2021.

The anchoring of objective inflation expectations to those of analyst surveys is very important as its omission provides inflation expectations that rise too sharply in the latter part of the estimated period, consequently leading to an unjustified decrease in the inflation risk premium.

Our research question is to identify the determinants to changes in breakeven inflation, an issue at the center of the economic policy debate since the inception of the ECB unconventional monetary policy measures. Our findings can be a first step to move forward and analyze important issues from a policy perspective. To this end, future research should be devoted to assessing whether the rise in inflation expectations over the medium and long term observed since the second half of 2021 is due to the positive effect of anchoring inflation expectations to the ECB target or to a worrying upward decoupling between long-term and short-term expectations. In addition, policy makers and investors can use our findings to investigate which variables affect the inflation risk premium and inflation expectations.
References


Huber, P. J. (1967). The behavior of maximum likelihood estimates under nonstandard conditions.


### A Appendix

#### A.1 Derivation of the system for the market prices of risk

Given a generic diffusion process $dW_t = h(W_t)dt + f(W_t)dZ^p_{W,t}$, where $Z^p_{W,t}$ is a standard Brownian motion under the objective $P$-measure, the essentially affine market price of risk$^{27}$

$$\Lambda^W_t = \gamma^W_0 + \gamma^W_1 g(W_t)$$

defines the relationship between $Z^p_{W,t}$ and the corresponding Brownian motion process under the risk-neutral $Q$-measure $Z^Q_{W,t}$, i.e.

$$Z^Q_{W,t} = Z^p_{W,t} + \int_0^t \Lambda^W_s ds = Z^p_{W,t} + \int_0^t \gamma^W_0 + \gamma^W_1 g(W_s) ds$$

$^{27}$See Cheridito et al. (2007) for a review of different specifications of market prices of risk.
Note that, in case of a CIR process, we have $h(W_t) = k_0 + k_1 W_t$, $f(W_t) = \sigma \sqrt{W_t}$ and $g(W_t) = \sqrt{W_t}$. Assuming standard conditions to exclude arbitrage opportunities,\(^{28}\) the relationship between Brownian motions allows to obtain under the $Q$-measure the same dynamics followed by the state variable $W$ under the $P$-measure, by appropriately adjusting the drift parameters.\(^{29}\) Since model (7-10) has three state variables, i.e. $W = (X, Y, V)$, with different stochastic processes where $X$ links to $Y$, the market price of risk is defined by a system that links the dynamics under the two probability measures.

The dynamics under the risk-neutral $Q$-measure

$$
\begin{align*}
&d \begin{pmatrix} X \\ Y \\ V \end{pmatrix} = \begin{pmatrix} 0 \\ \alpha \\ \theta \end{pmatrix} dt + \begin{pmatrix} -\lambda & \lambda & 0 \\ 0 & -\beta & 0 \\ 0 & 0 & -\phi \end{pmatrix} \begin{pmatrix} X \\ Y \\ V \end{pmatrix} dt + \\
&\begin{pmatrix} \eta & 0 & 0 \\ 0 & s & 0 \\ 0 & 0 & \sigma \sqrt{V} \end{pmatrix} \begin{pmatrix} dZ^Q_X \\ dZ^Q_Y \\ dZ^Q_V \end{pmatrix}
\end{align*}
$$

can be written in terms of the dynamics under the objective $P$-measure

$$
\begin{align*}
&d \begin{pmatrix} X \\ Y \\ V \end{pmatrix} = \begin{pmatrix} 0 \\ \mu \\ \delta \end{pmatrix} dt + \begin{pmatrix} -\kappa & \kappa & 0 \\ 0 & -\xi & 0 \\ 0 & 0 & -\psi \end{pmatrix} \begin{pmatrix} X \\ Y \\ V \end{pmatrix} dt + \\
&\begin{pmatrix} \eta & 0 & 0 \\ 0 & s & 0 \\ 0 & 0 & \sigma \sqrt{V} \end{pmatrix} \begin{pmatrix} dZ^P_X \\ dZ^P_Y \\ dZ^P_V \end{pmatrix}
\end{align*}
$$

\(^{28}\)See Cheridito et al. (2007) for details.

\(^{29}\)See the Appendix in Cecchetti (2020) for a simple derivation of the link between the risk neutral and objective dynamics for some stochastic processes, given the assumption of a proper market price of risk.
as

\[
\begin{pmatrix}
\dot{X} \\
\dot{Y} \\
\dot{V}
\end{pmatrix} =
\begin{pmatrix}
0 \\
\theta \\
\alpha
\end{pmatrix}
dt + \begin{pmatrix}
-\lambda & \lambda & 0 \\
0 & -\beta & 0 \\
0 & 0 & -\phi
\end{pmatrix}
\begin{pmatrix}
X \\
Y \\
V
\end{pmatrix}
dt + \\
\begin{pmatrix}
\eta & 0 & 0 \\
0 & s & 0 \\
0 & 0 & \sigma \sqrt{V}
\end{pmatrix}
\begin{pmatrix}
dZ^P_X \\
dZ^P_Y \\
dZ^P_V
\end{pmatrix}
\begin{pmatrix}
\gamma_0^X & -\gamma_0^Y & 0 \\
0 & \gamma_1^Y & 0 \\
0 & 0 & \gamma_1^V / \sqrt{V}
\end{pmatrix}
\begin{pmatrix}
X \\
Y \\
V
\end{pmatrix}
dt + \\
\begin{pmatrix}
0 \\
\alpha + s \gamma_0^Y \\
\theta + \sigma \gamma_0^V
\end{pmatrix}
dt + \\
\begin{pmatrix}
-\lambda + \eta \gamma_1^X & \lambda - \eta \gamma_1^X & 0 \\
0 & -\beta + s \gamma_1^Y & 0 \\
0 & 0 & -\phi + \sigma \gamma_1^V
\end{pmatrix}
\begin{pmatrix}
X \\
Y \\
V
\end{pmatrix}
dt + \\
\begin{pmatrix}
\eta & 0 & 0 \\
0 & s & 0 \\
0 & 0 & \sigma \sqrt{V}
\end{pmatrix}
\begin{pmatrix}
dZ^P_X \\
dZ^P_Y \\
dZ^P_V
\end{pmatrix}
\]

and the mapping for the drift parameters between the two measures is

\[
\begin{align*}
\kappa &= \lambda - \eta \gamma_1^X \\
\mu &= \alpha + s \gamma_0^Y \\
\xi &= \beta - s \gamma_1^Y \\
\delta &= \theta + \sigma \gamma_0^V \\
\phi &= \phi - \sigma \gamma_1^V
\end{align*}
\]

Note that the Feller condition for the CIR process governing the dynamics of \( V \) must apply for the solution to be bounded below by zero. Note also that \((\gamma_0^X, \gamma_0^Y, \gamma_0^V)\) and \((\gamma_1^X, \gamma_1^Y, \gamma_1^V)\) define the differences between the drift terms for the processes \( X \), \( Y \) and \( V \) within the objective \( \mathbb{P} \)-measure and risk-neutral \( \mathbb{Q} \)-measure and allow the market to incorporate time-varying inflation risk premia into prices.

### A.2 Inflation swap pricing

From equations (7) and (11), the price index at time \( T \) can be written

\[
I_T / I_0 = \exp \left( \int_0^T X_s ds - \frac{1}{2} \int_0^T V_s ds + \int_0^T \sqrt{V_s} dZ_{i,s} \right)
\]

24
where we can set \( I_0 = 1 \) without loss of generality. The cash flow of an inflation swap is equal to \( I_T - (1 + i_T)^T \) and since the present value of the inflation swap is nil at inception, we can write

\[
E^Q \left[ \exp \left( - \int_0^T r_s ds \right) \left( I_T - (1 + i_T)^T \right) \right] = 0
\]

We define the instantaneous nominal rate equal to the sum of the instantaneous real rate and expected inflation, \( r_t = R_t + X_t \), and substituting \( r_t \) and \( I_T \) obtain

\[
E^Q \left[ e^{-\int_0^T R_s ds} \left\{ E^Q \left[ e^{-\int_0^T X_s ds} e^{\int_0^T r_s ds - \frac{1}{2} \int_0^T V_s ds + \frac{1}{2} \int_0^T \sqrt{V} dZ_{1,s}} \right] - E^Q \left[ e^{-\int_0^T X_s ds} (1 + i_T)^T \right] \right\} \right] = 0
\]

which implies

\[
(1 + i_T)^T = \frac{E^Q \left[ \exp \left( - \frac{1}{2} \int_0^T V_s ds + \int_0^T \sqrt{V} dZ_{1,s} \right) \right]}{E^Q \left[ \exp \left( - \int_0^T X_s ds \right) \right]} = \frac{1}{E^Q \left[ \exp \left( - \int_0^T X_s ds \right) \right]} \quad (A.1)
\]

If we set \( E^Q \left[ \exp \left( - \int_0^T X_s ds \right) \right] = H(X, Y, \tau) \) where \( \tau = T - t \), \( H \) satisfies the following PDE

\[
\frac{1}{2} \eta^2 H_{XX} + \frac{1}{2} s^2 H_{YY} + \lambda(Y - X) H_X + (\alpha - \beta Y) H_Y - XH - \frac{\partial H}{\partial \tau} = 0
\]

We guess a solution of the form \( H = \exp(A(\tau) + B(\tau)X + C(\tau)Y) \) such that \( H_{XX} = B^2 H, H_{YY} = C^2 H, H_X = BH, H_Y = CH \) and obtain

\[
\frac{\partial B(\tau)}{\partial \tau} = -\lambda B(\tau) - 1
\]

\[
\frac{\partial C(\tau)}{\partial \tau} = \lambda B(\tau) - \beta C(\tau)
\]

\[
\frac{\partial A(\tau)}{\partial \tau} = \frac{1}{2} \eta^2 B^2(\tau) + \frac{1}{2} s^2 C^2(\tau) + \alpha C(\tau)
\]

The equation are solved by the use of an integrating factor and direct integration. We substitute the solutions into the expression for \( H(X, Y, T) \) into equation (A.1) and evaluate at \( \tau = T \). This gives equation (1).
A.3 Term structure of inflation swaps

The vectors $A(T), B(T), C(T)$ have the following expression under the $Q$-measure:

$$A(T) = -\frac{1}{T} \cdot \left[ \frac{\alpha \lambda}{\beta - \lambda} \left( \frac{1}{\beta} \left( T - \frac{1}{\beta}(1 - e^{-\beta T}) \right) - \frac{1}{\lambda} \left( T - \frac{1}{\lambda}(1 - e^{-\lambda T}) \right) \right) + \frac{s^2 \lambda^2}{2(\lambda - \beta)^2} \left( \frac{1}{\beta^2} \left( T - \frac{2}{\beta}(1 - e^{-\beta T}) \right) - \frac{1}{\beta^2} \left( T - \frac{1}{\lambda}(1 - e^{-2\beta T}) \right) \right) - \frac{2}{\beta \lambda} \left( T - \frac{1}{\beta}(1 - e^{-\beta T}) - \frac{1}{\lambda}(1 - e^{-\lambda T}) + \frac{1}{\beta + \lambda}(1 - e^{-(\beta + \lambda)T}) \right) \right]$$

$$B(T) = -\frac{1}{T} \cdot \frac{-(1 - e^{-\lambda T})}{\lambda}$$

$$C(T) = -\frac{1}{T} \cdot \frac{\lambda}{\beta - \lambda} \left( \frac{1}{\beta}(1 - e^{-\beta T}) - \frac{1}{\lambda}(1 - e^{-\lambda T}) \right)$$

(A.3)

The last four lines of equation (A.2) can be substituted by $\sigma_w^2(T)$ defined below by equation (A.6).

A.4 Distribution of inflation

Under the $Q^*$-measure the two members of equation (11), $w_T$ and $u_T$, have the following distribution. $w_T \sim N(\mu_w(T), \sigma_w^2(T))$ where

$$\mu_w(T) = (1 + it)^T - \frac{1}{2} \sigma_w^2(T)$$

(A.5)

and variance

$$\sigma_w^2(T) = \frac{s^2 \lambda^2}{(\lambda - \beta)^2} \left( \frac{1}{\beta^2} \left( T - \frac{2}{\beta}(1 - e^{-\beta T}) + \frac{1}{2\beta}(1 - e^{-2\beta T}) \right) \right)$$

(A.6)

$$- \frac{2}{\beta \lambda} \left( T - \frac{1}{\beta}(1 - e^{-\beta T}) + \frac{1}{\lambda}(1 - e^{-\lambda T}) + \frac{1}{\beta + \lambda}(1 - e^{-(\beta + \lambda)T}) \right)$$

$$+ \frac{1}{\lambda^2} \left( T - \frac{2}{\lambda}(1 - e^{-\lambda T}) + \frac{1}{2\lambda}(1 - e^{-2\lambda T}) \right)$$

$$+ \frac{s^2 \lambda^2}{2(\lambda - \beta)^2} \left( \frac{1}{\beta^2} \left( T - \frac{2}{\beta}(1 - e^{-\beta T}) \right) - \frac{1}{\beta^2} \left( T - \frac{1}{\lambda}(1 - e^{-2\beta T}) \right) \right)$$

This term corresponds to $G(T)$ in equation (2) of Section 2.

The distribution of $u_T$ is obtained from

$$du = -\frac{1}{2} V dt + \sqrt{V} dZ_t$$

$$dV = (\theta - \phi V) dt + \sigma \sqrt{V} dZ_V$$
that is a special case of the Heston (1993) model with characteristic function \( E[e^{\kappa u_T}] \) of \( u_T \) given by
\[
\exp(L(T) + M(T)V) \tag{A.7}
\]
where
\[
\begin{align*}
L(T) &= \frac{\theta(\phi + \gamma)}{2\sigma^2} T + \frac{2\theta}{\sigma^2} \ln \left( \frac{1 - k_0}{1 - k_0 e^{-\gamma T}} \right) \\
M(T) &= \frac{\phi + \gamma}{\sigma^2} \cdot \frac{1 - e^{\gamma T}}{1 - k_0 e^{\gamma T}}
\end{align*}
\]
and where
\[
\begin{align*}
\gamma &= \sqrt{\sigma^2(\zeta^2 + i\zeta) + \phi^2} \\
k_0 &= (\phi + \gamma)/(\phi - \gamma)
\end{align*}
\]
Then, the density function of \( u_T \), obtained by inverting the characteristic function, is given by
\[
h(u_T) = \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-i\zeta u_T} \exp(L(T) + M(T)V) d\zeta \tag{A.8}
\]
The cumulants of \( u_T \) are obtained in closed form by repeatedly differentiating the log of the characteristic function (A.7) with respect to the argument \( \zeta \) and evaluating the derivatives at \( \zeta = 0 \). The first cumulant of \( u_T \), i.e. the mean of \( u_T \), is given by
\[
\mu_u(T) = -\frac{1}{2} \left( V - \frac{\theta}{\phi} \right) \frac{1}{\phi} (1 - e^{-\phi T}) - \frac{1}{2\phi} T \tag{A.9}
\]
The second cumulant of \( u_T \), the variance of \( u_T \), is given by
\[
\begin{align*}
\sigma_u^2(T) &= \left( 1 + \frac{\sigma^2}{4\phi^2} \right) \left( -\frac{\theta}{\phi^2}(1 - e^{-\phi T}) + \frac{\theta}{\phi} T \right) \\
&\quad - \frac{\sigma^2}{2\phi^2} \left( -\frac{\theta}{\phi} e^{-\phi T} + \frac{\theta}{\phi^2} (1 - e^{-\phi T}) \right) \\
&\quad + \frac{\sigma^2}{4\phi^2} \left( -\frac{\theta}{\phi^2}(e^{-\phi} - e^{-2\phi T}) + \frac{\theta}{2\phi^2} (1 - e^{-2\phi T}) \right) \\
&\quad + \left( \left( 1 + \frac{\sigma^2}{4\phi^2} \right) \frac{1}{\phi} (1 - e^{-\phi T}) - \frac{\sigma^2}{2\phi^2} e^{-\phi T} + \frac{\sigma^2}{4\phi^2} \frac{1}{\phi} (e^{-\phi T} - e^{-2\phi T}) \right) V.
\end{align*}
\]
This term corresponds to \( H(T) + U(t) \cdot V \) in equation (2) of Section 2.

The distribution of inflation is defined in terms of the joint density of \( u_T \) and that of \( w_T \), that given their independence is the product of their marginals. The cumulants of the distribution of the logarithm of inflation are equal to the sum of the cumulants of \( u_T \) and that of \( w_T \). So basically to get the inflation variance we simply sum \( \sigma_u^2(T) \) and \( \sigma_w^2(T) \). This is why the two terms \( G(T) \) and \( H(T) + U(t) \cdot V \) in equation (2) of Section 2 can be separated.

27
A.5 Gram-Charlier expansion

Denote the first and second cumulant of inflation as $c_1 = \mu_w(T) + \mu_u(T)$ – defined in equations (A.5,A.9) – and $c_2 = \sigma^2_w(T) + \sigma^2_u(T)$ – defined in equations (A.6,A.10) – see Chateau and Dufresne (2017) for references. Standardize inflation and define it $x$. The first three Hermite polynomials $He_n$, for $n = 1, 2, 3$, are given by

$$
He_0 = 1 \\
He_1 = x \\
He_2 = x^2 - 1
$$

By the definition of Gram-Charlier expansion, the density of $x$ can be approximated up to the second order by

$$
f(x,T) \approx \left[ He_0 + c_1 He_1 + \frac{1}{2} \left( c_1^2 + c_2 - 1 \right) He_2 \right] \cdot n(x) \\
= \left[ 1 + c_1 x + \frac{1}{2} \left( c_1^2 + c_2 - 1 \right) (x^2 - 1) \right] \cdot n(x)
$$

where $n(\cdot)$ is the density function of the standard normal. In order to overcome the usual problem of having negative values of the term in square brackets we use the following normalization

$$
f(x,T) \approx \left[ 1 + c_1 x + \frac{1}{2} \left( c_1^2 + c_2 - 1 \right) (x^2 - 1) \right]^2 \cdot n(x)
$$

The probabilities of observing an inflation below zero and above 2% are equal to

$$
Pr(x < 0) = \int_{-\infty}^{0} f(x,T) dx \\
Pr(x > 0.02) = \int_{0.02}^{\infty} f(x,T) dx.
$$

Figure 12 reports $f(x,T)$ for $T = 2, 3, 5, 10$. Figure 13 reports the probabilities (A.11-A.12) for $T = 3$.

A.6 Pricing of caps and floors

The no-arbitrage price of a cap option is given by the expectation under the risk-neutral measure $\mathbb{Q}$ of the discounted payoff:

$$
\mathbb{E}^\mathbb{Q} \left[ D(T) \cdot \max(0, I_T - (1 + K)^T) \right].
$$
Applying the change-of-numeraire technique, this expectation can be computed as the product of the convenient numeraire, \( D(T) \), and the expectation under the \( Q^* \) forward measure:

\[
D(T) \cdot \mathbb{E}^{Q^*} \left[ \max(0, I_T - (1 + K)^T) \right]
\]

where

\[
D(T) = \mathbb{E}^Q \left[ e^{-\int_0^T X_s \, ds} \right] = (1 + i)^{-T}
\]

and the expectation can be written as

\[
\mathbb{E}^{Q^*} \left[ \max(0, (1 + i T)^T - (1 + K)^T) \right] = \mathbb{E}^{Q^*} \left[ \max(0, e^{u_{T+w_T}} - (1 + K)^T) \right]
\]

as function of \( u_T \) and \( w_T \) whose joint density is \( f(w_T, u_T) = f(w_T) \cdot f(u_T) \) by the independence between \( w_T \) and \( u_T \). By direct integration:

\[
\mathbb{E}^{Q^*} \left[ \max(0, e^{u_{T+w_T}} - (1 + K)^T) \right] = \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} \max \left(0, e^{u_{T+w_T}} - (1 + K)^T\right) f_{w_T} f_{u_T} \, dw_T \, du_T .
\]

As \( e^{u_{T+w_T}} > (1 + K)^T \) if and only if \( w_T > T \ln(1 + K) - u_T \), the expectation can be written as

\[
\int_{-\infty}^{+\infty} \int_{T \ln(1+K) - u_T}^{+\infty} e^{u_{T+w_T}} f_{w_T} f_{u_T} \, dw_T \, du_T - (1 + K) \int_{-\infty}^{+\infty} \int_{T \ln(1+K) - u_T}^{+\infty} f_{w_T} f_{u_T} \, dw_T \, du_T .
\]

(A.13)

- Under the \( Q^* \) measure \( w_T \sim N(\ln((1 + i T)^T) - \sigma_w^2/2, \sigma_w^2) \) with density,

\[
f_W = \frac{1}{\sqrt{2\pi}\sigma_w} e^{-\frac{1}{2} \left( \frac{w_T - \ln((1 + i T)^T) + \sigma_w^2/2}{\sigma_w^2} \right)^2}
\]

and distribution function

\[
F_W(w) = N \left( \frac{w - \ln((1 + i T)^T) + \sigma_w^2/2}{\sqrt{G}} \right)
\]

- The first term of (A.13) can be written as

\[
\int_{-\infty}^{+\infty} \frac{e^{u_T}}{\int_{\ln(1+K)^T}^{+\infty} e^{u_{T+w_T}} f_{w_T} \, dw_T} \, du_T .
\]

(A.14)

Because \( \int_{-\infty}^{+\infty} e^{u_{T+w_T}} f_{w_T} \, dw_T = \mathbb{E}^{Q^*} \left[ e^{u_T} \right] = 1 \), this term converges. The term in \( w \) can be written as

\[
\int_{\ln(1+K)^T}^{+\infty} e^{u_{T+w_T}} f_{w_T} \, dw_T \]

(A.15)
Since \( w_T \) is normal, \( e^{w_T} \) is lognormal with mean
\[
\mathbb{E}[e^{w_T}] = e^{\ln((1+iT)^T) - \sigma_w^2/2 + \sigma_w^2/2}
\]
variance
\[
\text{Var}[e^{w_T}] = (e^{\sigma_w^2} - 1) e^{2\ln((1+iT)^T) - \sigma_w^2/2 + \sigma_w^2}
\]
and density
\[
\frac{1}{\sqrt{\sigma_w^2 2\pi e^w}} e^{-\frac{1}{2} \left( \frac{(\ln(e^{w_T} - \ln(1+iT)^T) + \sigma_w^2/2)^2}{\sigma_w^2} \right)}
\]
equation (A.15) becomes
\[
\mathbb{E}[Q_1^{w_T} 1_{e^{w_T} > (1+K)^T}] = \int_{(1+i)^T}^{\infty} \frac{1}{\sqrt{2\pi \sigma_w^2}} e^{-\frac{1}{2} \left( \frac{(\ln(e^{w_T} - \ln(1+iT)^T) + \sigma_w^2/2)^2}{\sigma_w^2} \right)} d(e^{w_T}) .
\]
By a change of variable \( w = \ln(e^w) \), \( d(e^w) = e^w dw \)
\[
\int_{T \ln(1+k) - u_T}^{\infty} e^{w_T} \frac{1}{\sqrt{2\pi \sigma_w^2}} e^{-\frac{1}{2} \left( \frac{(\ln(e^{w_T} - \ln(1+iT)^T) + \sigma_w^2/2)^2}{\sigma_w^2} \right)} dW_T . \quad (A.16)
\]
Combining terms and completing the square, the exponent in (A.16) becomes:
\[
-\frac{1}{2\sigma_w^2}(w^2 + \ln((1+iT)^T)^2 + \sigma_w^4 - 2w \ln((1+iT)^T))
\]
\[
+ w\sigma_w^2 - \ln((1+iT)^T)\sigma_w^2 - 2\sigma_w^2 w
\]
\[
= -\frac{1}{2\sigma_w^2}(w - (\ln((1+iT)^T) + \sigma_w^2/2))^2 + \ln((1+iT)^T)
\]
and defining \( a = \frac{u_T - T \ln(1+K) + \ln((1+iT)^T) + \sigma_w^2/2}{\sigma_w^2} \), (A.16) can be written as
\[
(1+iT)^T \int_{T \ln(1+k) - u_T}^{\infty} \frac{1}{\sqrt{2\pi \sigma_w^2}} e^{-\frac{1}{2} \left( \frac{(u_T - T \ln(1+K) + \ln((1+iT)^T) + \sigma_w^2/2)^2}{\sigma_w^2} \right)} dW_T = (1+iT)^T \cdot N(a) .
\]
- The second term of (A.13) can be written as
\[
-(1+K)^T \int_{-\infty}^{+\infty} f_{u_T} [1 - F_W(T \ln(1+K) - u_T)] du_T
\]
\[
-(1+K)^T \int_{-\infty}^{+\infty} N(a - \sqrt{\sigma_w^2}) f_{u_T} du_T .
\]
where \([1 - F_W(T \ln(1+K) - u_T)] = N(a - \sqrt{\sigma_w^2})\).
- Combining the terms, the price of the cap (A.13) is given by
\[
(1+iT)^T \int_{-\infty}^{+\infty} ((1+iT)^T \cdot N(a) \cdot e^{w_T} - (1+K)^T \cdot N(a - \sqrt{\sigma_w^2})) \cdot f_{u_T} du_T
\]
The price of the floor option can be derived in a specular way.
For a different derivation, see also appendix B in Fleckenstein et al. (2017).
B Figures and Tables

Figure 1: Swap price

The Figure shows the convergence of the Monte Carlo swap price to the closed form swap price for a given choice of initial values of the state variables and maturity.
The Figure shows the convergence of the call price obtained with Monte Carlo simulation (black line) to the call price obtained with closed formula (blue line), for a given choice of initial values of the state variables and maturity.
Figure 3: Put price

Put option price: MonteCarlo exercise with Q-star measure - closed form

# paths = 3000
Discretization step = 1/365

The Figure shows the convergence of the put price obtained with Monte Carlo simulation (black line) to the put price obtained with closed formula (blue line), for a given choice of initial values of the state variables and maturity.
Table 1: Results of estimates

<table>
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<th>parameter</th>
<th>std.err.</th>
<th>t-stat</th>
<th>p-value</th>
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<td>$s$</td>
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<td>0.0000231</td>
<td>570.94</td>
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The Table reports the parameters of model (1, 2, 12) estimated using a quasi-Newton algorithm. The standard error (std.err.), the t-statistics (t-stat) and the p-value (p-value) are computed with the Huber sandwich estimator.
Table 2: pricing error of inflation swaps

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<th>std</th>
<th>RMSE</th>
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The Table reports the pricing error in basis points (error), the standard deviation (std.dev) and the root mean squared errors (RMSE) in basis points of inflation swaps in equation (1).

Table 3: pricing error of implied variances

<table>
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<th>std.dev.</th>
<th>RMSE</th>
</tr>
</thead>
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The Table reports the pricing errors in basis points (error), the standard deviation (std.dev) and the root mean squared errors (RMSE) of implied volatility in equation (2).
The Figure reports the observed and fitted inflation swaps under the $Q$-measure.
The Figure reports the observed and fitted standard deviation under the Q-measure. The observed standard deviation is calculated as the second moment of the density implied in inflation caps and floors quotes.

The Figure reports the factor loadings for the inflation swaps and for the implied variance. The loading of vectors $G$ and $H$ are multiplied by 100.
Figure 7: instantaneous and long-term inflation

The Figure reports the instantaneous inflation (X) and long-term inflation (Y) estimated from the model.
Figure 8: implied variance under the $Q$ and $P$ measure

The Figure reports the fitted implied variance under the $Q$ and the $P$ measure.
Figure 9: fitted $Q$-measure inflation swap (breakeven), fitted $P$-measure inflation swap (expected) and inflation risk premium

The Figure reports the fitted $Q$-measure inflation swap (breakeven), the fitted $P$-measure inflation swap (expected) and the inflation risk premium for the maturity 1-, 3-, 5-, 10-, and 5-year five year forward.
Figure 10: inflation risk premium

The Figure reports the average inflation risk premium and the 5\% and 95\% percentiles.
Figure 11: fitted Q-measure inflation, fitted P-measure inflation, inflation risk premium and expected inflation surveyed by the SPF.

The Figure reports the 1-year forward fitted P-measure (breakeven) inflation, the fitted Q-measure (expected) inflation, the inflation risk premium, SPF annual inflation expected after one year (left panel) and the mean of the aggregate probability distribution of SPF annual inflation expected after 4 years (right panel).
Figure 12: $\mathbb{P}$-measure inflation densities

The Figure reports the densities of inflation under the $\mathbb{P}$ measure for the 2, 3, 5 and 10 year maturity.
Figure 13: probability of inflation lower than 0 and greater than 2% over the following 3 years

The Figure reports the $P$-measure ($\cdots$) and $Q$-measure ($-\hphantom{-}$) probability that inflation is lower than 0% and greater than 2% on average over the following three years.

Figure 14: 5-year five year forward expected inflation and inflation risk premium for different models

The Figure reports 5-year 5 year forward expected inflation and the inflation risk premium estimated by this paper (CGP), that estimated as in Adrian et al. (2013) (ACM), Joslin et al. (2011) (JSZ) and Pericoli (2019) (PER). JSZ and PER are at monthly frequency. The average of the long-term expected inflation estimated with the ACM and JSZ models is imposed to be equal to 1.9%, the long-term inflation aim of the ECB. The long-term expected inflation estimated with the CGP and PER models is unconstrained.
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